# Computational Physics with Maxima or R: Example 3, Time Independent Schroedinger's Equation in 1D * 

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This document is example3.pdf which applies some of the
topics in the third chapter of a series of notes titled
Computational Physics with Maxima or R , and is made available
via the author's webpage http://www.csulb.edu/~woollett/
to encourage the use of the Maxima and R languages for computational
physics projects of modest size.
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to others as long as you charge no more than the costs of printing.
Code files which accompany Example 3 are
1. FW.mac finite rectangular well
2. FW.R
3. LJ6-12.mac Lennard Jones 6/12 potential
4. LJ6-12.R
The code files from Chapter 3, cp3.mac and cp3.R, are
    also needed to run the examples.
A very useful reference is the new textbook:
Computational Physics: A Practical Introduction to Computational Physics
            and Scientific Computing, Athens, 2014,
            by Konstantinos Anagnostopoulos.
Free copies in various formats are available on the webpage:
    http://www.physics.ntua.gr/~konstant/ComputationalPhysics/
The author's webpage is:
http://www.physics.ntua.gr/~}konstan
```

Feedback from readers is the best way for this series of notes to become more helpful to users of $\mathbf{R}$ and Maxima. All comments and suggestions for improvements will be appreciated and carefully considered.

## 1 Transforming Schroedinger's Time Independent Equation to Dimensionless Form

We assume a scalar particle (spin zero). Using ordinary units, the wave function $\psi(x)$, for one dimensional problems, is a solution of the eigenvalue equation

$$
\begin{equation*}
\frac{\hbar^{2}}{2 m} \frac{d^{2}}{d x^{2}} \psi(x)+(E-V(x)) \psi(x)=0 \tag{1.1}
\end{equation*}
$$

in which $E$ is the energy of the particle in ergs, $V(x)$ is the potential energy in ergs, and $x$ is a length in centimeters. The coordinate space wave function $\psi(x)$ is assumed to be normalized according to (integrating over the region in which $\psi(x)$ is nonzero)

$$
\begin{equation*}
\int|\psi(x)|^{2} d x=1 \tag{1.2}
\end{equation*}
$$

since $d x|\psi(x)|^{2}$ is (in the absence of any other information) the probability that the particle described by $\psi(x)$ will be found in the interval $(x, x+d x)$, and the sum of the probabilities of mutually exclusive outcomes must add up to unity.

For Schroedinger's time independent equation in 1D, we can usually take $\psi(x)$ to be a real function of $x$, and from (1.2) we see that the dimension of $\psi(x)$ is $1 / \sqrt{c m}$.. Quoting Landau and Lifshitz, Quantum Mechanics, third revised edition reprinted 2003, p. 55,

Schroedinger's equation for the wave functions $\psi$ of stationary states is real, as are the conditions imposed on its solution. Hence its solutions can always be taken as real (These assertions are not valid for systems in a magnetic field; it is assumed that the potential energy does not depend explicitly on the time: the system is either closed or in a constant (non-magnetic) field).

The eigenfunctions of non-degenerate values of the energy are automatically real, apart from an unimportant phase factor. ... The wave functions corresponding to the same degenerate energy level need not be real, but by a suitable choice of linear combinations of them we can always obtain a set of real functions.

We assume there exists a quantity $L$ with the dimensions of cm ., chosen for convenience, in terms of which we can define a dimensionless coordinate $\tilde{x}$

$$
\begin{equation*}
\tilde{x}=x / L \tag{1.3}
\end{equation*}
$$

We also assume there exists an energy $V_{0}$ (with dimensions of ergs) associated with the potential energy in terms of which we can define a dimensionless potential energy $\tilde{V}(\tilde{x})$

$$
\begin{equation*}
\tilde{V}(\tilde{x})=\frac{V(x)}{V_{0}} \tag{1.4}
\end{equation*}
$$

and a dimensionless energy $\tilde{E}$

$$
\begin{equation*}
\tilde{E}=\frac{E}{V_{0}} \tag{1.5}
\end{equation*}
$$

We also define a dimensionless coordinate space wave function $\tilde{\psi}(\tilde{x})$

$$
\begin{equation*}
\tilde{\psi}(\tilde{x})=\sqrt{L} \psi(x) \tag{1.6}
\end{equation*}
$$

in terms of which we have the transformed Schroedinger's equation

$$
\begin{equation*}
\frac{d^{2} \tilde{\psi}(\tilde{x})}{d \tilde{x}^{2}}+\gamma^{2}(\tilde{E}-\tilde{V}(\tilde{x})) \tilde{\psi}(\tilde{x})=0 \tag{1.7}
\end{equation*}
$$

where

$$
\begin{equation*}
\gamma=\sqrt{\frac{2 m L^{2} V_{0}}{\hbar^{2}}} \tag{1.8}
\end{equation*}
$$

and we have a normalization condition in terms of $\tilde{x}$ and $\tilde{\psi}(\tilde{x})$ :

$$
\begin{equation*}
\int \tilde{\psi}(\tilde{x})^{2} d \tilde{x}=1 \tag{1.9}
\end{equation*}
$$

The "uncertainty relation" is a condition on the product of the uncertainty $\Delta x$ in the position of the particle, and the uncertainty $\Delta p_{x}$ in the simultaneous $x$ component of the mechanical momentum of the particle:

$$
\begin{equation*}
\Delta x \Delta p_{x} \geq \frac{\hbar}{2} \tag{1.10}
\end{equation*}
$$

in which, for a property $A$,

$$
\begin{equation*}
\Delta A=\sqrt{(\Delta A)^{2}} \tag{1.11}
\end{equation*}
$$

and

$$
\begin{equation*}
\left.(\Delta A)^{2}=\left\langle(A-<A>)^{2}\right\rangle=\left\langle A^{2}\right\rangle-<A\right\rangle^{2} . \tag{1.12}
\end{equation*}
$$

We define a dimensionless $x$ component of momentum $\tilde{p}_{x}$

$$
\begin{equation*}
\tilde{p}_{x}=\frac{L}{\hbar} p_{x} \tag{1.13}
\end{equation*}
$$

in terms of which (1.10) becomes

$$
\begin{equation*}
\Delta \tilde{x} \Delta \tilde{p}_{x} \geq \frac{1}{2} . \tag{1.14}
\end{equation*}
$$

The expectation value of $p_{x}^{n}$

$$
\begin{equation*}
\left\langle p_{x}^{n}\right\rangle=\int \psi^{*}(x)\left(\frac{\hbar}{i} \frac{d}{d x}\right)^{n} \psi(x) d x \tag{1.15}
\end{equation*}
$$

becomes

$$
\begin{equation*}
\left\langle\tilde{p}_{x}^{n}\right\rangle=\int \tilde{\psi}^{*}(\tilde{x})\left(\frac{1}{i} \frac{d}{d \tilde{x}}\right)^{n} \tilde{\psi}(\tilde{x}) d \tilde{x} \tag{1.16}
\end{equation*}
$$

When we can assume $\psi(x)$ is real (most of the time), $\left\langle p_{x}\right\rangle$ is either zero or a pure imaginary number, since

$$
\begin{equation*}
<p_{x}>=\int \psi(x) \frac{\hbar}{i} \frac{d \psi(x)}{d x} d x \tag{1.17}
\end{equation*}
$$

But $<p_{x}>$ must be real, and thus we conclude it must also be zero. For consistency, this implies that

$$
\begin{equation*}
\int \psi(x) \frac{d \psi(x)}{d x} d x=0 . \tag{1.18}
\end{equation*}
$$

## 2 The Finite Rectangular Potential Well: Energy Levels and Wave Functions

We assume a finite well such that $V(x)=V_{0}$ for $x \leq 0$ and also for $x \geq L>0$, while $V(x)=0$ for $0<x<L$. Transforming to dimensionless units as described in the previous section, we then have $\tilde{V}(\tilde{x})=1$ for $\tilde{x} \leq 0$ and for $\tilde{x} \geq 1$, and $\tilde{V}(\tilde{x})=0$ for $0<\tilde{x}<1$. In the following, we drop the tildes, with $\tilde{x} \rightarrow x, \tilde{V}(\tilde{x}) \rightarrow V(x), \tilde{E} \rightarrow E, \tilde{p}_{x} \rightarrow p_{x}$, and $\tilde{\psi}(\tilde{x}) \rightarrow y(x)$, so we are seeking energy eigenvalues $E$ and the associated energy eigenfunctions $y(x)$ such that $y(x)$ is a real continuous function satisfying the equation

$$
\begin{equation*}
\frac{d^{2}}{d x^{2}} y(x)+\gamma^{2}(E-V(x)) y(x)=0 \tag{2.1}
\end{equation*}
$$

and such that $y(x) \rightarrow 0$ as $x \rightarrow \pm \infty$ in such a way that we can satisfy the normalization condition

$$
\begin{equation*}
\int_{-\infty}^{\infty} y(x)^{2} d x=1 \tag{2.2}
\end{equation*}
$$

and we also satisfy the basic uncertainty relation $\Delta x \Delta p_{x} \geq \frac{1}{2}$.

Using R graphics methods, we can make a simple plot of the dimensionless rectangular well (which now looks like a finite square well), and add a hypothetical energy level (in red).

```
> setwd("c:/k3")
> source("cp3.R")
> plot(0, type="n",xlim=c(-2,2),ylim=c(-2,2),xlab="x",ylab="V(x)")
> lines(c(-2,0), c(1,1), lwd = 3,col = "blue" )
> lines(c(0,0), c(0,1), lwd = 3, col = "blue" )
> lines(c(0,1),c(0,0), lwd = 3, col = "blue" )
> lines(c(1,1),c(0,1), lwd = 3, col = "blue" )
> lines( c(1,2), c(1,1), lwd = 3, col = "blue")
> mygrid()
> lines( c(0,1),c(0.3,0.3), lwd = 3, col = "red" )
```

which produces the plot


Figure 1: Dimensionless Finite Well

### 2.1 Finite Well Analytic Solution

We first seek an analytic solution of the finite well problem. In the regions in which $V(x)=1$, the solutions of (2.1) which vanish for large values of $|x|$ are

$$
\begin{equation*}
y(x)=B_{1} e^{k_{1} x}, \quad x \leq 0 \tag{2.3}
\end{equation*}
$$

and

$$
\begin{equation*}
y(x)=A_{2} e^{-k_{1} x}, \quad x \geq 1, \tag{2.4}
\end{equation*}
$$

in which

$$
\begin{equation*}
k_{1}=\gamma \sqrt{1-E}, \tag{2.5}
\end{equation*}
$$

and $B_{1}$ and $A_{2}$ are constants.
The general solution in the region in which $V(x)=0$ can be written in the form

$$
\begin{equation*}
y(x)=A \sin (k x+\delta), \quad 0 \leq x \leq 1, \tag{2.6}
\end{equation*}
$$

in which

$$
\begin{equation*}
k=\gamma \sqrt{E} . \tag{2.7}
\end{equation*}
$$

This last equation can be solved for $E$ :

$$
\begin{equation*}
E=\frac{k^{2}}{\gamma^{2}} \tag{2.8}
\end{equation*}
$$

We can also then express $k_{1}$ in terms of $k$ :

$$
\begin{equation*}
k_{1}=\sqrt{\gamma^{2}-k^{2}} \tag{2.9}
\end{equation*}
$$

The required continuity of both $y$ and $y^{\prime}$ implies the continuity of the ratio $y^{\prime} / y$. We have

$$
\begin{equation*}
\frac{y^{\prime}(x)}{y(x)}=k_{1}, \quad x \leq 0 \tag{2.10}
\end{equation*}
$$

and

$$
\begin{equation*}
\frac{y^{\prime}(x)}{y(x)}=-k_{1}, \quad x \geq 1 \tag{2.11}
\end{equation*}
$$

and

$$
\begin{equation*}
\frac{y^{\prime}(x)}{y(x)}=k \cot (k x+\delta), \quad 0 \leq x \leq 1 . \tag{2.12}
\end{equation*}
$$

Then the continuity of $y^{\prime} / y$ at $x=0$ implies

$$
\begin{equation*}
\tan (\delta)=\frac{k}{k_{1}}=\frac{k}{\sqrt{\gamma^{2}-k^{2}}} \tag{2.13}
\end{equation*}
$$

And the continuity of $y^{\prime} / y$ at $x=1$ implies

$$
\begin{equation*}
\tan (k+\delta)=-\frac{k}{k_{1}}=-\tan (\delta) \tag{2.14}
\end{equation*}
$$

We expand the left hand side of this last equation, using the identity

$$
\begin{equation*}
\tan (A+B)=\frac{\tan (A)+\tan (B)}{1-\tan (A) \tan (B)} \tag{2.15}
\end{equation*}
$$

and use again (2.13) to get

$$
\begin{equation*}
k=\frac{\left(2 k^{2}-\gamma^{2}\right)}{2 \sqrt{\gamma^{2}-k^{2}}} \tan (k) \tag{2.16}
\end{equation*}
$$

This equation will involve real numbers provided we have

$$
\begin{equation*}
0<k<\gamma \tag{2.17}
\end{equation*}
$$

We can search for values of $k$ which satisfy both (2.16) and (2.17), which will then imply corresponding energy eigenvalues using (2.8). A graphical search can be achieved by plotting the left and right hand sides of (2.16) on the same plot and looking for intersections.

### 2.1.1 Analytic Energies and Wave Functions using Maxima

Our function kroot_plot (kkmin, kkmax, ymn, ymx ) (available for use once the code file $\mathbf{F W}$.mac is loaded) is designed to partially automate such a graphical search.

```
Frhs(k):= (float ( (2*k^2 - gam2) *tan(k)/2/sqrt (gam2 - k^2)) ) $ 
```

After loading FW.mac, gam2 is a global parameter which stands for $\gamma^{2}$, and in this code file we use $\gamma=50$.

```
(%i1) load(cp3);
(%o1) "c:/k3/cp3.mac"
(%i2) load(FW);
    gam = 50 gam2 = 2500
h = 0.01, xdecay = 0.5 , ypleft = 1.0E-8 ypright = 1.0E-8
(%O2) "c:/k3/FW.mac"
(%i3) kroot_plot (1,49,0,60) $
plot2d: some values were clipped.
```

which produces the plot


Figure 2: Graphical Search for k Roots
Placing the cursor over the intersections of the curves, we see that k roots occur approximately at values $\mathrm{k}=3.01,6.03,9.06, \ldots$
$\mathbf{F a ( k )}$ is a function, also defined in $\mathbf{F W}$.mac, which is zero at these special values of $\mathbf{k}$. Also defined is $\mathbf{k t o E}(\mathbf{k})$ which converts $\mathbf{k}$ to $\mathbf{E}$.

```
Fa(k):=(float (k - (2*k^2 - gam2)*tan(k)/2/sqrt (gam2 - k^2)) )$
ktoE (kv) := (kv^2/gam2) $
```

This function $\mathbf{F a}(\mathbf{k})$ (of one variable) can then be directly used with the core Maxima function find_root to find the ground state energy $\mathbf{E 0}$.

```
(%i4) Fa(2.9);
(%04) -3.2290071
(%i5) Fa(3.1);
(%o5) 2.0655923
(%i6) k0 : find_root(Fa,2.9,3.1);
(%06) 3.0206914
(%i7) EO : ktoE(k0);
(%०7) 0.00364983
(%i8) plot_analytic(E0) $
    E = 0.00364983
x_mean = 0.5
delx = 0.18802
ydy_sum = -1.78676518E-16
delp = 2.9619274 delx*delp = 0.556902
number of nodes = 0
```

which produces the plot of the normalized analytic ground state wave function with zero nodes.


Figure 3: Analytic Normalized Ground State Wave Function
The function plot_analytic (E) (see below) calls analytic_wf(E) (see code file Fw.mac), and the latter file calculates delx, delp, $\mathbf{x}$ _mean, and $\mathbf{y d y}$ _sum. The calculated value of $\mathbf{y} d \mathbf{y}$ _sum corresponds to the integral (1.18) which should be zero.

The function find_root also appears to find a spurious root at or near $\pi / 2$ which is not a physical solution. The function

$$
\begin{equation*}
F_{a}(k)=k-\frac{\left(2 k^{2}-\gamma^{2}\right)}{2 \sqrt{\gamma^{2}-k^{2}}} \tan (k) . \tag{2.18}
\end{equation*}
$$

is not a continuous function at $k=n \frac{\pi}{2}, \quad n=1,3,5, \ldots$ where $\tan (k)$ is not continuous, and find_root cannot be trusted at such points. For example,

```
(%i9) Fa(1.55);
(%09) 1201.7787
(%i10) Fa(1.59);
(%010) -1298.109
(%i11) k0s : find_root(Fa,1.55,1.59);
(%o11) 1.5707963
(%i12) Fa(k0s);
(%o12) 4.07689764E+17
(%i13) EOs : ktoE(k0s);
(%O13) 9.8696044E-4
(%i14) plot_analytic(E0s) $
    E = 9.8696044E-4
x_mean = 0.700405
delx = 0.212493
ydy_sum = 2.22044605E-16
delp = 6.6877608*(-1)^0.5 delx*delp = 1.4210999*(-1)^0.5
number of nodes = 0
```

which produces the plot of a discontinuous zero node wave function:


Figure 4: Discontinuous Spurious Root Wave Function
Note also the calculation produced a purely imaginary value for delp. Note also the large value of $\mathrm{Fa}(\mathrm{kOs})$, which should be a very small number close to a physical root.

The function analytic_wf(E) creates global analytic normalized wave functions yn1 (x), yn2(x), yn0(x), and yna ( $\mathbf{x}$ ) ( and also computes and prints analytic values of $\Delta x$, represented by delx, and $\Delta p$, represented by delp). $\mathrm{yn} 1(\mathbf{x})$ is the wave function for $x<0, \mathrm{yn} 2(\mathbf{x})$ is the wave function for $x>1, \mathrm{ynO}(\mathbf{x})$ is the wave function for $0 \leq x \leq 1$. yna ( $\mathbf{x}$ ) is the wave function for all $x$, usable by plot 2 d , but not by integrate, created by the line

```
yna(x) :=
    ( if x < 0 then ynl(x)
        else if x > 1 then yn2(x)
        else ynO(x) ),
```

Ignoring normalization, one can write a continuous $y_{u}(x)$ as

$$
\begin{aligned}
y_{u}(x) & =\sin (\delta) e^{k_{1} x}, \quad x \leq 0, \quad k_{1}=\gamma \sqrt{1-E} \\
& =\sin (k x+\delta), \quad 0 \leq x \leq 1, \quad k=\gamma \sqrt{E} \\
& =\sin (k+\delta) e^{k_{1}} e^{-k_{1} x}, \quad x \geq 1
\end{aligned}
$$

Normalization then requires calculating

$$
\begin{equation*}
D=\int_{-\infty}^{\infty} y_{u}^{2}(x) d x \tag{2.19}
\end{equation*}
$$

and a normalized wave function $y_{n}(x)$ is then

$$
\begin{equation*}
y_{n}(x)=\frac{y_{u}(x)}{\sqrt{D}} \tag{2.20}
\end{equation*}
$$

The function plot_analytic(E) used above calls analytic_wf(E) and nodes_analytic (ddx), prints out the number of nodes, and makes a plot of $\mathbf{y n a}(\mathbf{x})$.

```
plot_analytic(E) :=
block ( [ddx : 0.001, xvL, ynL, ymn, ymx,
        xmn: -0.25, xmx : 1.25, numer], numer:true,
    analytic_wf(E),
    xvL : makelist (x,x,0,1,ddx),
    ynL : map(yn0, xvL),
    ymn : floor( lmin(ynL) ),
```

```
ymx : 1 + floor( lmax (ynL) ),
print(" number of nodes = ", nodes_analytic(ddx) ),
plot2d(yna(x), [x,xmn,xmx], [ylabel, "yna(x)"], [y,ymn,ymx],
    [style,[lines,2]],[gnuplot_preamble,"set grid"]))$
```

The function nodes_analytic (dx) counts the number of nodes implied by the function yn0 (x) created by analytic_wf(E). The count is naturally restricted to the region $0 \leq x \leq 1$ and the argument $\mathbf{d x}$ is the step size used.

```
nodes_analytic ( dx ) :=
block( [num:0, xv:0, xnew, f1, f2, numer], numer:true,
    do (
            f1 : yn0 (xv),
            xnew : xv + dx,
            if xnew > 1 then return(),
            f2 : ynO(xv + dx),
            if f1*f2 < O then num : num + 1,
            xv : xnew),
            num) $
```

A function levels_analytic (kmax) returns a list of analytic eigen energies related to $k$ via $k=\gamma \sqrt{E}$. The corresponding eigenvalues of $k$ are separated by roughly $d k=3$ and lie in the middle of the intervals $\left[n_{1} \frac{\pi}{2}, n_{2} \frac{\pi}{2}\right.$ ], where $n_{1}$ and $n_{2}$ are adjacent odd integers with $n_{2}>n_{1}$. The ground state (zero node soln) corresponds to $k=3.02, n_{1}=1, n_{2}=3$.

```
levels_analytic(kmax) :=
block( [ka,kb,kv,level:0, nmax, Elist : [], numer], numer:true,
    nmax : ceiling(2*kmax/%pi),
    print(" nodes E "),
    print( " "),
    /* make nmax an odd integer */
    if evenp (nmax) then nmax : nmax + 1,
    /* analytic kv using n*pi/2 + 0.1 with n odd */
    for j:1 step 2 thru nmax do (
        [ka, kb] : bracket_basic( Fa, j*%pi/2 + 0.1, 0.01, 0.005),
        kv : find_root( Fa, ka, kb),
        Ev : ktoE(kv),
        Elist : cons(Ev, Elist),
        print( " ", level, " ", Ev ),
        level : level + 1),
    reverse(Elist) ) $
```

Here is an example of the use of levels_analytic:

```
(%i15) levels_analytic(20);
    nodes E
        0.00364983
        0.0145973
        0.032836
        0.0583552
        0.0911389
        0.131165
        0.178405
(%o15) [0.00364983,0.0145973,0.032836,0.0583552,0.0911389,0.131165,0.178405]
(%i16) E4 : %[5];
(%o16) 0.0911389
(%i17) plot_analytic(E4)$
    E = 0.0911389
x_mean = 0.5
delx = 0.297122
ydy_sum = -8.32667268E-17
delp = 14.787571 delx*delp = 4.3937127
number of nodes = 4
```

which produces the plot


Figure 5: Analytic Four Node Wave Function $E=0.0911389$
The function levels_analytic calls bracket_basic (func, $\mathbf{x}, \mathrm{dx}, \mathbf{x a c c}$ ) which looks for a sign change in func, starting with $\mathbf{x}$, and increasing $\mathbf{x}$ by $d \mathbf{x}$ each step. If a sign change is found, then we back up to the previous $\mathbf{x}$ and search with new $d \mathbf{x}$ value one half of the previous value.

```
bracket_basic(func, xx,dxx, xacc) :=
block([ x:xx, dx:dxx, x1,x2,it:0,itmax:1000],
    do (
        it : it + 1,
        if it > itmax then (
        print(" can't find change in sign "),
            return([0, 0 ])),
        x1 : x,
        x2 : x + dx,
        if debug then print(" it = ",it," x1 = ", x1," x2 = ", x2," dx = ", dx),
        if func(x1)*func(x2) < 0 then (
            if abs(dx) < xacc then return([x1,x2]),
                x : x - dx,
                dx : dx/2)
        else x : x2))$
```

Here is an example using bracket_basic with func = sin:

```
(%i18) [xa,xb] : bracket_basic(sin,3,0.01,0.001);
(%o18) [3.14125,3.141875]
(%i19) xv : find_root(sin,xa,xb);
(%019) 3.1415927
```

Here is an example of locating the zero node ground state value of $\mathbf{k}$ for the finite potential well problem using func $=\mathbf{F a}$ (see (2.18)).

```
(%i20) [ka,kb] : bracket_basic(Fa,1.6,0.1,0.05);
(%o20) [3.0,3.025]
(%i21) kv : find_root(Fa,ka,kb);
(%o21) 3.0206914
```


### 2.1.2 Analytic Energies and Wave Functions using R

We can search for values of $k$ which satisfy both (2.16) and (2.17), which will then imply corresponding energy eigenvalues using (2.8). A graphical search can be achieved by plotting the left and right hand sides of (2.16) on the same plot and looking for intersections. Our function kroot_plot (kkmin, kkmax , ymn , ymx ) is designed to partially automate such a graphical search.

```
Frhs = function (k) { (2*k^2 - gam2)*tan(k)/2/sqrt(gam2 - k^2) }
kroot_plot = function (kkmin,kkmax, ymn, ymx) {
    curve (Frhs,kkmin,kkmax, n=200, col = "red",lwd = 3,ylim = c (ymn,ymx),
                        xlab = "k", ylab = "")
    lines( c(kkmin,kkmax), c(kkmin,kkmax),col = "blue",lwd = 3 ) }
```

After loading FW. R, gam2 is a global parameter which stands for $\gamma^{2}$, and in this code file we use $\gamma=50$.

```
> source("cp3.R")
> source("FW.R")
    gam = 50 gam2 = 2500
h = 0.01, xdecay = 0.5 , ypleft = 1e-08 ypright = 1e-08
> kroot_plot(1,49,0,60)
> mygrid()
```

which produces the plot


Figure 6: Graphical Search for k Roots
Let's zoom in on the beginning of this plot and add a custom grid.

```
> kroot_plot(1, 10,0,10)
> abline( v = seq(1,10,by = 1),h = seq(0,10,by = 1) )
```

which produces the plot


Figure 7: Graphical Search for k Roots
Physically valid k roots occur roughly at values slightly larger than $k=3,6,9, \ldots$ The vertical red lines correspond to values of $k=\pi / 2,3 \pi / 2, \ldots$ where $\tan (k)$ both changes sign and has an arbitrarily large magnitude. The intersections with the vertical red lines are not physically valid roots, as we will see.

An function which should be zero at a physical root is called $\mathbf{F a}(\mathbf{k})$ and we also define $\mathbf{k t o E}(\mathbf{k})$, which converts $\mathbf{k}$ to $\mathbf{E}$.

```
Fa = function(k) { k - (2*k^2 - gam2)*tan(k)/2/sqrt (gam2 - k^2) }
ktoE = function (k) k^2/gam2
```

This function $\mathbf{F a}(\mathbf{k})$ (of one variable) can then be directly used with uniroot to find the ground state energy $\mathbf{E O}$.

```
> Fa(2.9)
[1] -3.22901
> Fa(3.1)
[1] 2.06559
> k0 = uniroot(Fa, c(2.9, 3.1), tol=1e-16)$root
> k0
[1] 3.02069
> EO = ktoE(kO)
> EO
[1] 0.00364983
> plot_analytic(EO)
    E = 0.00364983
x_mean = 0.5
delx = 0.18802
ydy_sum = -1.78677e-16
delp = 2.96193 delx*delp = 0.556902
ymn = 0 ymx = 2
number of nodes = 0
```

which produces the plot of the normalized analytic ground state wave function with zero nodes.


Figure 8: Analytic Normalized Ground State Wave Function
The function uniroot also appears to find a spurious root at or near $\pi / 2$ which is not a physical solution. The function

$$
\begin{equation*}
F_{a}(k)=k-\frac{\left(2 k^{2}-\gamma^{2}\right)}{2 \sqrt{\gamma^{2}-k^{2}}} \tan (k) . \tag{2.21}
\end{equation*}
$$

is not a continuous function at $k=n \frac{\pi}{2}, \quad n=1,3,5, \ldots$ where $\tan (k)$ is not continuous, and uniroot cannot be trusted at such points. For example,

```
> Fa(1.55)
[1] 1201.78
> Fa(1.59)
[1] -1298.11
> k0s = uniroot(Fa,c(1.55,1.59),tol=1e-16)$root
> kOs
[1] 1.5708
> Fa(kOs)
[1] -3.01869e+16
> EOs = ktoE (kOs)
> EOs
[1] 0.00098696
> plot_analytic(E0s)
    E = 0.00098696
    x_mean = 0.700405
    delx = 0.212493
    ydy_sum = 1.11022e-16
    delp = NaN delx*delp = NaN
    ymn = 0 ymx = 2
    number of nodes = 0
Warning message:
In sqrt(delp2) : NaNs produced
```

which produces the plot of a discontinuous zero node wave function:


Figure 9: Discontinuous Spurious Root Wave Function
Note also the calculation produced a purely imaginary value for delp, which $\mathbf{R}$ writes as NaN (not a number). Note also the large value of $\mathrm{Fa}(\mathrm{kOs})$, which should be a very small number close to a physical root.

The function analytic_wf(E) creates global analytic normalized wave functions $\mathbf{y n} 1(\mathbf{x}), \mathrm{yn} \mathbf{2 ( x )}, \mathrm{yn} \mathbf{0}(\mathrm{x})$, and yna ( $\mathbf{x}$ ) ( and also computes and prints analytic values of $\Delta x$, represented by delx, and $\Delta p$, represented by delp). $\mathrm{yn} 1(\mathbf{x})$ is the wave function for $x<0, \mathrm{yn} 2(\mathbf{x})$ is the wave function for $x>1, \mathrm{ynO}(\mathbf{x})$ is the wave function for $0 \leq x \leq 1$. yna ( $\mathbf{x}$ ) is the wave function for all $x$, usable by sapply, but not by curve, created by the line

```
yna <<- function (x) { if (x < 0) yn1(x) else if (x > 1) yn2(x) else yn0(x) }
```

Ignoring normalization, one can write a continuous $y_{u}(x)$ as

$$
\begin{aligned}
y_{u}(x) & =\sin (\delta) e^{k_{1} x}, \quad x \leq 0, \quad k_{1}=\gamma \sqrt{1-E} \\
& =\sin (k x+\delta), \quad 0 \leq x \leq 1, \quad k=\gamma \sqrt{E} \\
& =\sin (k+\delta) e^{k_{1}} e^{-k_{1} x}, \quad x \geq 1 .
\end{aligned}
$$

Normalization then requires calculating

$$
\begin{equation*}
D=\int_{-\infty}^{\infty} y_{u}^{2}(x) d x \tag{2.22}
\end{equation*}
$$

and a normalized wave function $y_{n}(x)$ is then

$$
\begin{equation*}
y_{n}(x)=\frac{y_{u}(x)}{\sqrt{D}} \tag{2.23}
\end{equation*}
$$

The function plot_analytic(E) used above calls analytic_wf(E) and nodes_analytic (ddx), prints out the number of nodes, and makes a plot of $\mathbf{y n a}(\mathbf{x})$.

```
plot_analytic = function (E) {
    ddx = 0.001
    xmn = - 0.25
    xmx = 1.25
```

```
analytic_wf(E)
xL = seq(xmn,xmx ,by = ddx)
yL = sapply(xL, yna)
ymn = floor( min(yL) )
ymx = 1 + floor( max (yL) )
cat (" ymn = ",ymn," ymx = ",ymx, "\n")
cat (" number of nodes = ", nodes_analytic(ddx), "\n" )
plot(xL, yL, type = "l", lwd = 3, col = "blue", xlab = "x",
    ylab = "ya(x)",tck=1, ylim = c(ymn,ymx) ) }
```

The function nodes_analytic (dx), used above, counts the number of nodes implied by the function yn0 ( $\mathbf{x}$ ) created by analytic_wf(E). This function looks in the region $0 \leq x \leq 1$, using intervals of size $\mathbf{d x}$.

```
nodes_analytic = function ( dx ) {
    num = 0
    xv = 0
    repeat {
        f1 = yn0(xv)
        xnew = xv + dx
        if (xnew > 1) break
        f2 = ynO(xv + dx)
        if (f1*f2 < 0 ) num = num + 1
        xv = xnew }
    num }
```

A function levels_analytic (kmax) returns a list of analytic energy eigenvalues related to $k$ via $k=\gamma \sqrt{E}$. The corresponding eigenvalues of $k$ are separated by roughly $d k=3$ and lie in the middle of the intervals [ $n_{1} \frac{\pi}{2}, n_{2} \frac{\pi}{2}$ ], where $n_{1}$ and $n_{2}$ are adjacent odd integers with $n_{2}>n_{1}$. The ground state solution corresponds to $k=3.02, n_{1}=1, n_{2}=3$.

```
levels_analytic = function (kmax) {
    mmax = 20
    EL = rep (NA, rmax)
    level = 0
    nmx = ceiling(2*kmax/pi)
    ## make nmx an odd integer
    if ( is.even (nmx) ) nmx = nmx + 1
    cat (" nodes E \n ")
    cat ( " \n ")
## analytic kv using n*pi/2 + 0.1 with n odd
    for ( j in seq(1, nmx, by=2) ) {
        out = bracket_basic( Fa, j*pi/2 + 0.1, 0.01,0.005)
        kv = uniroot( Fa, out, tol = le-16) $root
        Ev = ktoE(kv)
        EL[ j ] = Ev
        cat ( " ", level, " ", Ev, "\n" )
        level = level + 1 }
    EL[!is.na(EL)] }
```

Here is an example of the use of levels_analytic:

```
> EL = levels_analytic(20)
    nodes E
    0.00364983
    0.0145973
        0.032836
        0.0583552
        0.0911389
        0.131165
        0.178405
```

```
> EL
[1] 0.00364983 0.01459726 0.03283598 0.05835517 0.09113890 0.13116525 0.17840502
> E4 = EL[5]
> E4
[1] 0.0911389
> plot_analytic(E4)
    E = 0.0911389
x_mean = 0.5
delx = 0.297122
ydy_sum = -8.32667e-17
delp = 14.7876 delx*delp = 4.39371
ymn = -2 ymx = 2
number of nodes =4
```

which produces the plot


Figure 10: Analytic Four Node Wave Function $E=0.0911389$
The function levels_analytic calls bracket_basic (func, $\mathbf{x}, \mathrm{dx}, \mathbf{x a c c}$ ) which looks for a sign change in func, starting with $\mathbf{x}$, and increasing $\mathbf{x}$ by $d \mathbf{x}$ each step. If a sign change is found, then we back up to the previous $\mathbf{x}$ and search with new $d \mathbf{x}$ value one half of the previous value. Note that debug = FALSE is set when loading the file FW.R.

```
bracket_basic \(=\) function (func, \(\mathbf{x x}, \mathrm{dxx}, \mathbf{x a c c}\) ) \{
        \(\mathbf{x}=\mathbf{x x}\)
        \(\mathrm{dx}=\mathrm{dx} \mathbf{x}\)
        it \(=0\)
        itmax \(=1000\)
        anerror \(=\) FALSE
    repeat \{
        it = it + 1
        if (it > itmax) \{
        cat (" can't find change in sign \(\backslash n\) ")
                anerror = TRUE
                break
        \(\mathrm{x} 1=\mathrm{x}\)
        \(x 2=x+d x\)
```




Here is an example using bracket_basic with func = sin:

```
> out = bracket_basic(sin,3,0.01,0.001)
> out
[1] 3.14125 3.14187
> uniroot(sin,out,tol=1e-16) $root
[1] 3.14159
```

Here is an example of locating the zero node ground state value of $\mathbf{k}$ for the finite potential well problem using func $=\mathbf{F a}$ (see (2.21)).

```
> out = bracket_basic(Fa,1.6,0.1,0.05)
> out
[1] 3.000 3.025
> kv = uniroot(Fa,out,tol = 1e-16) $root
> kv
[1] 3.02069
> Ev = ktoE(kv)
> Ev
[1] 0.00364983
```


### 2.2 Numerical Runge Kutta Finite Well Solution

It is easiest to use Runge-Kutta methods for a case in which the potential $V(x)$ is discontinuous at one or more values of $x$. In this example $V(x)$ is discontinuous at $x=0$ and $x=1$. The Runge-Kutta method automatically supplies the first derivatives at the spatial grid points $\mathbf{x L}$.

### 2.2.1 Numerical Energies and Wave Functions using Maxima

We use our homemade $\mathbf{r k} \mathbf{4}$ routine for the Runge-Kutta integration. When the file $\mathbf{F w}$. mac is loaded, a number of global parameters are defined. The top the the file FW.mac has the lines:

```
/* FW.mac uses Runge-Kutta
    for finite well.
    dimensionless units
    v = 1 for x < 0 and x > 1
    v = 0 for 0 < x < 1
    y'\prime}(x)+\operatorname{gam}2*(E-V(x))*y(x) = 
    gam2 = gam^2 = 2500
    gam = 50 = sqrt (2*m*L^2*VO/ hbar^2)
*/
/* initial global parameters: */
( N : 100,
    h : 0.01,
    gam : 50,
    gam2 : gam^2,
    xdecay : 0.5, /* start yL1 integration at x = -xdecay */
        /* start yR integration at x = 1 + xdecay */
    ypleft : 1e-8,
    ypright : 1e-8,
    print(" gam = ",gam, " gam2 = ", gam2),
    print(" h = ", h, ", xdecay = ", xdecay, ", ypleft = ",
                                    ypleft," ypright = ", ypright))$
```

We integrate from a point $\mathbf{x}=-\mathbf{x d e c a y}$ chosen so that we can assume $\mathbf{y}(-\mathbf{x d e c a y})=\mathbf{0}$ to the point $x=0$, thus defining a grid $\mathbf{x L 1}$ of integration points, a list $\mathbf{y} \mathrm{L} 1$ of values of $y(x)$ at these grid points, and a list $\mathbf{y p L} 1$ of values of $y^{\prime}(x)$ at these grid points where $V=1$.

We assume an arbitrary small value ypleft for the first derivative $y^{\prime}$ at this starting point. The resulting wave function, the solution of a homogeneous equation, can be later normalized, which will, in effect, amount to choosing the correct initial first derivative at the left starting point.

The final value of $y$ and $y^{\prime}$ thus generated become the initial values of $y$ and $y^{\prime}$ for integration through the region where $V=0,0 \leq x \leq 1$, thus generating a grid $\mathbf{x L 2}$ of integration points, a list $\mathbf{y} \mathbf{L} 2$ of values of $y(x)$ at these grid points, and a list ypL2 of values of $y^{\prime}(x)$ at these grid points where $V=0$.

The integration in the region $x>1$ is done by starting at a location $\mathbf{x}=1+\mathbf{x d e c a y}$ where we can assume $y=0$ and we again assign an arbitrary (but negative) first derivative - ypright. We then integrate toward smaller values of $x$ until we reach $x=1$. Since we are hence integrating in the direction in which the physical solution is growing, we avoid integration instability problems produced by small roundoff and integration algorithm errors.

We then multiply the lists $\mathbf{y L 1}, \mathbf{y p L 1}, \mathbf{y L 2}$, and $\mathbf{y p L 2}$ by a factor which assures us that the final value of $y(x)$ produced by the independent rightward and leftward integrations agree at the matching point $x=1$. The value of $y(x)$ can be made to agree at the matching point for any energy $E$. However, the resulting wave function values will still be discontinuous
because the first derivatives will not agree at the matching point.
The crucial step, then, is to design a function $F(E)$, say, that is zero (to within numerical errors) when the derivatives agree at the matching point. We can then look for the locations of sign changes in $F(E)$ to locate the energy eigenvalues.

The first step needed, in order to be able to design such a function $\mathbf{F}(\mathbf{E})$, is to design a function $\mathbf{w f}(\mathbf{E})$ which uses Runge Kutta methods to find a un-normalized wave function corresponding to a given total energy $\mathbf{E}$. Here is our code for such a wave function integrator, as listed in FW.mac.

```
/* wf(E) creates ** un-normalized ** numerical wave functions
    using Runge-Kutta routine rk4.
    The wave functions are stored in global xL1, yL1,ypL1, xL2, yL2, ypL2, xR, yR, ypR .
        Program also defines **global** nleft, nright, ncenter
        the global xL1 grid extends from -xdecay to 0 and
        the global xL2 grid extends from 0 to 1 and
        the global xR grid extends from 1 to 1 + xdecay
*/
wf(E) :=
block([ glr,gc, outL, fac, numer], numer : true,
    if (E < O) or (E > 1) then (
        print(" need 0 < E < 1 "),
        return(false)),
    ncenter : N,
    if not integerp(ncenter) then (
        print (" ncenter = ",ncenter," is not an integer"),
        return(false)),
    nleft : round(xdecay/h),
    nright : nleft,
    if wfdebug then print(" nleft = ",nleft," ncenter = ",ncenter," nright = ",nright),
    glr : gam2*(E - 1), /* g(x) for x < 0 and x > 1 */
    gc : gam2*E, /* g(x) for 0<x< < */
    if wfdebug then print(" glr = ",glr," gc = ", gc),
    /* construct xL1, yL1, and ypL1 for -xdecay < x < 0 */
    outL : rk4(['y2, - glr* 'y1], ['y1, 'y2], [ 0, ypleft], ['x, -xdecay, 0, h] ),
    xL1 : take(outL,1),
    yL1 : take(outL,2),
    ypL1 : take(outL, 3),
    /* construct xL2, yL2, and ypL2 for 0 < x < 1 */
    outL : rk4(['y2, - gc* 'y1], ['y1, 'y2], [ last(yL1), last(ypL1)], ['x, 0, 1, h] ),
    xL2 : take(outL,1),
    yL2 : take(outL,2),
    ypL2 : take(outL, 3),
/* construct xR, yR, and ypR for 1 < x < 1 + xdecay */
    outL : rk4(['y2, - glr* 'y1], ['y1, 'y2], [ 0, -ypright], ['x, 1 + xdecay, 1, -h] ),
    xR : take (outL, 1),
    YR : take (outL, 2),
    YpR : take(outL, 3),
    xR : reverse(xR),
    yR : reverse(yR),
    ypR : reverse(ypR),
    if wfdebug then print(" yR(1) = ", first(yR)),
```

```
fac : first(yR)/last(yL2),
if wfdebug then print(" fac = ",fac),
yL1 : fac*yL1,
ypL1 : fac*ypL1,
yL2 : fac*yL2,
ypL2 : fac*ypL2,
done) $
```

The second step needed to design $\mathbf{F}(\mathbf{E})$ is to create a function dy_diff() which returns a normalized difference of the first derivatives $y_{L}^{\prime}(x)-y_{R}^{\prime}(x)$ evaluated at the matching point $x=1$. We return this difference divided by the value of $y(x=1)$.

```
/* dy_diff() uses global ypL2, ypR, and yR,
    returns a normalized difference of derivatives
        (YI' (1) - YR'(1)/ YR(1)
*/
dy_diff() :=
block([dy_left, dy_right, numer], numer:true,
    dy_left : last(ypL2),
    dy_right : first(ypR),
    (dy_left - dy_right)/abs(first(yR)) ) $
```

For example,

```
(%i1) load(cp3);
(%o1) "c:/k3/cp3.mac"
(%i2) load(FW);
    gam = 50 gam2 = 2500
h = 0.01, xdecay = 0.5, ypleft = 1.0E-8 ypright = 1.0E-8
(%O2) "c:/k3/FW.mac"
(%i3) wf(0.5);
(%03) done
(%i4) dy_diff();
(%04) 35.071714
(%i5) last(ypL2);
(%O5) -0.00190471
(%i6) first(ypR);
(%06) -0.237432
(%i7) first(yR);
(%O7) 0.00671558
```

Here, finally, is our code for $\boldsymbol{F}(\mathrm{E})$ :

```
/* energy eigenvalue if global function F(E) = 0 .
    F(E) calls wf(E) then returns dy_diff(), but
        returns false if E > 0.
    */
F(E) :=
block( [ numer], numer:true,
    if E < O or E > 1 then (
        print(" in F(E), E = ",E," should be between 0 and 1 "),
        return(false)),
    wf(E),
    dy_diff()) $
```

And here is an example of using $\mathbf{F}(\mathbf{E})$ to produce a rough graphical survey of the possibilities:

```
(%i8) EL : makelist(e,e,0.1,0.9,0.01)$
(%i9) FL : map(F, EL) $
(%i10) plot2d([discrete,EL,FL],[xlabel,"E"],[Ylabel,"F(E)"])$
```

which produces the plot


Figure 11: Crude Plot of $\mathrm{F}(\mathrm{E})$
We can use $\mathbf{F}(\mathbf{E}$ ) to search for energy eigenvalue candidates. We do this with a function bracket (func, $\mathbf{x}, \mathbf{d x}, \mathbf{x a c c}$ ) which attempts to return two values of $\mathbf{x}$ at which func has the opposite sign.

```
/* bracket is a modified version of bracket_basic, designed to work with
    the function F(E) or F1(k) which can return 'false'.
    bracket looks for a sign change in func,
    starting with xx, and increasing xx by dxx each step.
    If sign change is found, then we back up to the previous xx
    and search with new dxx value one half of the previous value.
    normally returns [ea,eb] or [ka,kb], but if can't find change in sign,
    then returns [0,0], and if func returns false, then
    bracket returns false.
*/
bracket (func, xx,dxx, xacc) :=
block([f1,f2, x:xx, dx:dxx, xx1, xx2,it:0,itmax:1000],
    do (
        it : it + 1,
        if debug then print(it),
        if it > itmax then (
            print(" can't find change in sign "),
            return([0, 0 ])),
        xx1 : x,
    xx2 : x + dx,
    if debug then print(" it = ",it," xx1 = ", xx1," xx2 = ", xx2," dx = ", dx),
    f1 : func(xx1),
    if not f1 then (
            print(" in bracket, f1 = false , xx1 = ", xx1, " dx = ", dx),
            return(f1)),
    f2 : func(xx2),
```

```
if not f2 then (
    print(" in bracket, f2 = false , xx2 = ", xx2, " dx = ", dx),
    return(f2)),
if debug then print (" f1 = ",f1," f2 = ",f2),
if f1*f2 < O then (
    if abs(dx) < xacc then return([xx1,xx2]),
    x : x - dx,
    dx : dx/2)
else x : xx2) )$
```

Here is an example of using bracket with the function $\mathbf{F}(\mathbf{E})$. This example produces the zero node ground state case, and we plot the un-normalized wave function pieces produced by $\mathbf{w f}(\mathrm{E})$.

```
(%i11) [ea,eb] : bracket(F,0.0005,0.0001,0.00005);
(%o11) [0.003625,0.00365]
(%i12) e : find_root(F,ea,eb);
(%o12) 0.00364983
(%i13) wf(e);
(%o13) done
(%i14) plot2d([[discrete,xL1,yL1],[discrete,xL2,yL2],[discrete, xR,yR]],
    [xlabel,"x"], [ylabel, "y(x)"], [legend,false])$
```

which produces a plot of the un-normalized ground state wave function with zero nodes.


Figure 12: Numerical Un-normalized Ground State Wave Function
We can check the normalized difference in slopes at the matching point for the solution produced by $\mathbf{w f}(\mathbf{E})$ :

```
(%i15) dy_diff();
(%o15) -2.49565076E-14
```

We can check the number of nodes with the function num_nodes ().

```
/* count the number of nodes in yL2 */
num_nodes() :=
block([ n, numer], numer:true,
    n : 0,
    for j thru (length(yL2) - 1) do
        if yL2[ j ] * yL2[ j + 1 ] < 0 then n : n + 1,
    n) $
```

For the numerical ground state solution generated above by $\mathbf{w f}(\mathbb{E})$ we get:

```
(%i16) num_nodes();
(%016) 0
```

A function normalize () uses the current wave function pieces produced by $\mathbf{w f}(\mathbf{E})$ and uses our Simpson's rule function simp to produce global normalized wave function (lists) $x$ n and $y$ n. normalize () uses our function merge (aL1, aL2) to combine the separate lists into one list. After calling normalize (), one can check the normalization:

```
(%i17) normalize()$
AA = 6652.6824
x_mean = 0.5
delx = 0.18802
(%i18) simp(xn,yn^2);
(%018) 1.0
```

Here is the code for normalize ().

```
/* normalize() uses the current global xL1,yL1, xL2,yL2, xR, yR and
    the utility functions merge and simp to define global
    xn and yn, with the latter being normalized.
    */
normalize() :=
block ( [AA,x_mean, x2_mean,delx,delx2, numer ], numer:true,
    xn : merge( xL1, merge( rest(xL2), rest(xR))),
    yn : merge( yL1, merge( rest(yL2), rest (yR))),
    /* we need xn to have odd # of elements to use simp */
    if evenp ( length (xn) ) then (
        xn : rest (xn),
        yn : rest (yn)),
    if debug then print ( " fll(xn) = ", fll(xn) ),
    if debug then print( " fll(yn) = ", fll(yn) ),
    AA : simp (xn,yn^2),
    print( " AA = ",AA),
    yn : yn/sqrt(AA),
    if debug then print( " fll(yn) = ", fll(yn) ),
    x_mean : simp(xn, xn * yn^2),
    print(" x_mean = ", x_mean),
    x2_mean : simp (xn, xn^2 * yn^2),
    delx2 : x2_mean - x_mean^2, /* this should be positive! */
    delx : sqrt(delx2),
    print(" delx = ", delx),
    done) $
```

Once we have used normalize (), we can use the function yn_plot_current (), which uses the current lists $\mathbf{x n}$ and yn.

```
(%i19) yn_plot_current()$
ymax = 1.3867012
```

which produces the plot


Figure 13: Numerical Normalized Ground State Wave Function
Here is our code for yn_plot_current ():

```
/* yn_plot_current() uses the currently defined normalized set (xn,yn)
*/
yn_plot_current() :=
block([ymn, ymx, numer],numer:true,
    ymn : float(floor( lmin(yn) )),
    ymx : float( 1 + floor( lmax (yn) ) ),
    print(" ymax = ", lmax(yn) ),
    plot2d( [discrete, xn, yn], ['y,ymn,ymx],
        [ylabel,"y"], [xlabel,"x"],
        [style,[lines,3]], [legend, false], [gnuplot_preamble,"set grid"]))$
```

The more versatile function $\mathbf{y n} \_\mathbf{p l o t}(\mathbf{E}, \mathbf{x m i n}, \mathbf{x m a x})$ does three tasks in succession, first calling $\mathbf{w f}(\mathbf{E})$ to create the un-normalized wave function pieces, then calling normalize () to create the normalized wave function lists $\mathbf{x n}$ and $\mathbf{y n}$, and finally making a plot of the normalized wave function, using $x \min$ and $x$ max to control the horizontal display.

Here is an example dealing with the first excited (one node) state.

```
(%i20) [ea,eb] : bracket(F,0.01,0.005,0.001);
(%O20) [0.014375,0.015]
(%i21) e : find_root(F,ea,eb);
(%O21) 0.0145973
(%i22) yn_plot (e, -0.5,1.5)$
E = 0.0145973
number of nodes = 1, dy_diff = -7.59213486E-14
AA = 1278.4068
x_mean = 0.5
delx = 0.276562
normalized ymax = 1.3865517
```

which produces the plot


Figure 14: Numerical Normalized First Excited State Wave Function
Here is our code for yn_plot:

```
/* yn_plot(E,xmin,xmax) first calls wf(E) to create
    un-normalized wave functions corresponding to the
        given energy E. Then normalizes those wave functions
        to produce the lists xn and yn. Finally makes a plot
        of yn over only the region (xmn, xmx) */
yn_plot(E,xmn,xmx) :=
block([ymn, ymx, numer], numer:true,
    wf(E),
    print(" E = ",E ),
    print(" number of nodes = ",num_nodes(),", dy_diff = ",dy_diff() ),
    normalize(),
    print(" normalized ymax = ", lmax(yn) ),
    ymn : floor( lmin(yn) ),
    ymx : 1 + floor( lmax (yn) ),
    plot2d( [discrete, xn, yn], ['x,xmn, xmx],
        ['y,ymn,ymx], [ylabel,"y"], [xlabel,"x"],
                [style, [lines, 3] ], [legend, false], [gnuplot_preamble,"set grid"]))$
```

We now want to construct a function levels (...) which will produce a list of the energy levels, found using our numerical Runge-Kutta methods, starting with the ground state energy, and continuing up to some maximum energy. Some experimentation shows that instead of using $\boldsymbol{F}(\mathbf{E})$ with a succession of small values of $\mathbf{E}$, it is easier to use a function $\mathbf{F 1}(\mathbf{k})$, since the $\mathbf{k}$ values corresponding to the energy eigenvalues are larger numbers of order $O(1)$. Here is such a function which we call F1 (k).

```
/* energy eigenvalue if global function F1(k) = 0 .
    F1(k) calls wf(E) then returns dy_diff(), but
        returns false if k <= 0 or >= gam.
    */
F1(k) :=
block( [ numer], numer:true,
    if k <= 0 or k >= gam then (
            print(" in F1(k), k = ",k," k should be greater than 0 and less than ",gam),
            return(false) ),
        wf(k^2/gam2),
        dy_diff())$
```

Here is an example of use of $\mathbf{F 1}(\mathbf{k})$ to find the ground state energy:

```
(%i23) F1(2.9);
(%023) 34.29503
(%i24) F1(3.05);
(%024) -49.882755
(%i25) [ka,kb] : bracket(F1,2.9,0.05,0.02);
(%O25) [3.0125,3.025]
(%i26) k0 : find_root(F1,ka,kb);
(%026) 3.0206914
(%i27) EO : ktoE(kO);
(%O27) 0.00364983
(%i28) wf(EO);
(%028) done
(%i29) dy_diff();
(%029) -1.49739046E-13
(%i30) num_nodes();
(%030) 0
```

which reveals a zero node wave function with a very small value of dy_diff(), a signal of a good wave function.
We can make a crude plot of $\mathbf{F 1 ( k )}$ versus $\mathbf{k}$

```
(%i31) kL : makelist(k,k,1,49,0.5)$
(%i32) F1L : map(F1, kL)$
(%i33) time(%);
(%033) [6.23]
(%i34) plot2d([discrete,kL,F1L],[xlabel,"k"],[ylabel,"F1(k)"])$
```

which produces the plot


Figure 15: Crude Plot of $\mathrm{F} 1(\mathrm{k})$ versus k
Let's use $\mathbf{F 1}(\mathrm{k})$ to look at the region $2.8<=\mathrm{k}<=3.3$ :

```
(%i315 kL : makelist(k,k,2.8,3.3,0.05)$
(%i36) F1L : map(F1, kL) $
(%i37) plot2d([discrete,kL,F1L],[xlabel,"k"],[ylabel,"F1(k)"])$
```

which produces the plot


Figure 16: F1(k) Zoom
Placing the cursor on the $\mathbf{F 1}=0$ intersections shows roots at roughly $\mathbf{k}=3.01$ and $\mathbf{k}=3.06$. The first root is close to the valid ground state case as shown above. We can use bracket with $\mathbf{F 1}(\mathrm{k})$ to refine the second root.

```
(%i38) [ka,kb] : bracket(F1,3.03,0.01,0.005);
(%038) [3.0775,3.08]
(%i39) kv : find_root(F1,ka,kb);
(%o39) 3.0799546
(%i40) Ev : ktoE(kv);
(%O40) 0.00379445
(%i41) wf(Ev);
(%041) done
(%i42) num_nodes();
(%042) 0
(%i43) dy_diff();
(%O43) -1.60000886E+16
```

The large value of dy_diff() shows that this second, slightly larger root, is an un-physical solution. Since we already have a valid zero node solution at the lower energy, there cannot be a second zero node solution at another, higher, energy. This pattern persists, with the physical root being smaller, and the un-physical root being slightly larger. This pattern provides the rationale for our code for levels (...). Once we have found a solution with a given number of nodes, we reject all solutions with higher energy but the same number of nodes. The unphysical roots correspond to a sudden change in which (left or right) integration function has the larger slope magnitude at the matching point.

Here is our code for levels (kmin, kmax, dk, kacc ).

```
/* levels(kmin,kmax,dk, kacc ) returns a list [Ea, Eb,...] of energy levels with
    increasingly larger number of nodes in energy range (Emin, Emax)
    according to Emin = kmin^2/gam^2, and Emax = kmax^2/gam^2.
    uses F1(k) (inside bracket) to find roots, and calls wf(E), num_nodes() and
        dy_diff() for each root found,
    Uses bracket and find_root.
    The arguments (dk, kacc) are used to call bracket, and do not describe
        the accuracy of the energy levels found.
    Once a good energy e.v. is found we look for the region of
        energies with one more node and search there.
    Includes an interactive continue or stop input.
    Searching for the k eigenvalues via F1(k) is easier than searching
        directly for the E eigenvalues via F(E) for the case gam = 50 which we
        consider in our examples.
*/
```

```
levels(kmin, kmax, dk, kacc ) :=
block([ k,knext, kroot, eroot, eL, ka, kb, nn, nlast : -1, r, numer], numer:true,
    k : kmin,
    el : [ ], /* list eL will hold energy eigenvalues found */
    do (
            if k > kmax then return(), /* exit do loop */
            print("---------------- levels ------------------------
            print(" nlast = ", nlast),
            print(" kstart = ", k," dk = ", dk ),
            [ka, kb] : bracket(F1,k, dk, kacc),
            print(" ka = ",ka," kb = ",kb),
            if float(ka) = 0.0 then (
                print(" can't find bracket interval "),
                    print(" k = ",k),
                return() ),
            kroot : find_root(F1, ka, kb),
            print(" kroot = ", kroot),
            eroot : kroot`2/gam2,
            print(" eroot = ", eroot),
            wf (eroot),
            nn : num_nodes(),
            print(" number of nodes = ", nn),
            print(" dy_diff at x = 1 is ", dy_diff() ),
            eL : cons(eroot, eL),
            nlast : nn,
            r : read (" input c; or s; "),
            if string(r) = "s" then return(), /* exit do loop */
    /* search for a k value greater than kb which produces
                a wave function with nn + 1 nodes */
            knext : kb + dk,
        do (
            wf(knext`2/gam2),
            if num_nodes() > nlast then (
                k : knext,
                    return() )
            else knext : knext + dk)),
    reverse(eL) ) $
```

Here is an example of using levels. To continue to the next energy eigenvalue, one enters $\mathbf{c}$; at the prompt (for "continue"). Actually, any letter except $\mathbf{s}$; will cause the program to continue.

```
(%i44) EL : levels(1,8,0.05,0.02);
---------------- levels ---------------------
    nlast = -1
    kstart = 1 dk = 0.05
    ka = 3.0125 kb = 3.025
    kroot = 3.0206914
    eroot = 0.00364983
    number of nodes = 0
    dy_diff at x = 1 is -2.49565076E-14
    input c; or s;
C;
---------------- levels
    nlast = 0
    kstart = 3.125 dk = 0.05
    ka = 6.0375 kb = 6.05
    kroot = 6.040956
    eroot = 0.0145973
    number of nodes = 1
    dy_diff at x = 1 is
        2.18273877E-13
    input c; or s;
c;
```

```
-_--_--_--_------ levels _--------------------
    nlast = 1
    kstart = 6.2 dk = 0.05
    ka = 9.05 kb = 9.0625
    kroot = 9.0603555
    eroot = 0.032836
    number of nodes = 2
    dy_diff at x = 1 is 7.10273285E-14
    input c; or s;
C;
(%044) [0.00364983,0.0145973,0.032836]
```

We can then use $y n \_p l o t(E, x m i n, x m a x)$ to both construct the lists $\mathbf{x n}$ and $\mathbf{y n}$ of the normalized wave function and make a plot. We can also use makelist to construct one list xyn, say, which combines the lists $\mathbf{x n}$ and yn into one. In the following, we do not show the plots produced by the calls to yn_plot.

```
(%i45) yn_plot(EL[1],-0.5,1.5)$
E = 0.00364983
number of nodes = 0, dy_diff = -2.49565076E-14
AA = 6652.6824
x_mean = 0.5
delx = 0.18802
normalized ymax = 1.3867012
(%i46) xyn0 : makelist([xn[j],yn[j]],j,1,length(xn))$
(%i47) fll(xyn0);
(%047) [[-0.5,0.0],[1.5,0.0], 201]
```

We continue in this manner with the first excited state and the second excited state.

```
(%i48) yn_plot(EL[2],-0.5,1.5)$
E = 0.0145973
number of nodes = 1, dy_diff = -2.05936658E-12
AA = 1278.4068
x_mean = 0.5
delx = 0.276562
normalized ymax = 1.3865517
(%i49) xyn1 : makelist([xn[j],yn[j]],j,1,length(xn)) $
(%i50) yn_plot(EL[3],-0.5,1.5)$
E = 0.032836
number of nodes = 2 , dy_diff = -2.00652203E-12
AA = 365.3835
x_mean = 0.5
delx = 0.290108
normalized ymax = 1.385693
(%i51) xyn2 : makelist([xn[j],yn[j]],j,1,length(xn))$
```

We can then combine the plots for the wave functions of these three lowest lying states.
(\%i52) plot2d([[discrete, xyn0], [discrete, xyn1], [discrete, xyn2]],
[xlabel,"x"], [ylabel,"y"],[style, [lines,2]], [legend, "EO", "E1", "E2"]) \$
which produces the plot


Figure 17: Finite Well: Lowest Three Energy Level Wave Functions
Just for practice, we can write these three wave functions to a file in the local folder, and then restart Maxima, read in the file contents, and make a plot based on the file contents. We have discussed some details of such read and write actions within Maxima in Chapter 2 of Maxima by Example.

The simplest approach is to use save (filePath, a1,a2,a3,...), where objects a1, a2, etc are object names bound to quantities known to Maxima. The names and the objects bound to the names are stored in Lisp language format in the file requested (which is created if it does not yet exist, and overwritten if it already exists). You can, of course, open that file with a text editor to see the contents, written in Lisp.

One can use load to load in that file into a new Maxima session, and the names and objects will then be available for use in your new Maxima session. If you don't remember the names you used in your previous session, you can use values; to generate a list of currently known object names.

```
(%i53) EO : EL[1];
(%O53) 0.00364983
(%i54) E1 : EL[2];
(%O54) 0.0145973
(%i55) E2 : EL[3];
(%055) 0.032836
(%i56) save("c:/k3/fw1.dat",E0, xyn0,E1, xyn1,E2,xyn2);
(%056) "c:/k3/fw1.dat"
```

If we look at the top of the file c:/k3/fw1. dat with a text editor (such as Notepad ++ ) we see:

```
;;; -*- Mode: LISP; package:maxima; syntax:common-lisp; -*-
(in-package :maxima)
(DSKSETQ | $e0| 0.0036498306077588829)
(ADD2LNC '|$e0| $VALUES)
(DSKSETQ $XYNO
            '((MLIST SIMP) ((MLIST SIMP) -0.5 0.0)
            ((MLIST SIMP) -0.48999999999999999 1.2769296242295217E-12)
            ((MLIST SIMP) -0.47999999999999998 2.8785287263561221E-12)
            ((MLIST SIMP) -0.46999999999999997 5.2117339500682875E-12)
            ((MLIST SIMP) -0.46000000000000002 8.8694267352155839E-12)
            ((MLIST SIMP) -0.45000000000000001 1.4781088055950433E-11)
```

etc., etc.

We now restart Maxima and load in cp3.mac, FW.mac, and the data file created using save above.

```
(%i1) load(cp3);
(%o1) "c:/k3/cp3.mac"
(%i2) load(FW);
    gam = 50 gam2 = 2500
h=0.01, xdecay = 0.5, ypleft = 1.0E-8 ypright = 1.0E-8
(%02) "c:/k3/FW.mac"
(%i3) load("c:/k3/fw1.dat");
(%o3) "c:/k3/fw1.dat"
(%i4) values;
(%O4) [mydate,_binfo%,N,h,gam,gam2,xdecay,ypleft,ypright,E0, xyn0,E1, xyn1, E2, xyn2]
(%i5) EO;
(%O5) 0.00364983
(%i6) E1;
(%06) 0.0145973
(%i7) E2;
(%O7) 0.032836
(%i8) fll(xyn0);
(%08) [[-0.5,0.0], [1.5,0.0],201]
(%i9) head(xyn0);
(%09) [[-0.5,0.0],[-0.49,1.27692962E-12],[-0.48,2.87852873E-12],
    [-0.47,5.21173395E-12], [-0.46,8.86942674E-12], [-0.45,1.47810881E-11]]
(%i10) plot2d([[discrete,xyn0],[discrete,xyn1],[discrete,xyn2]],
    [xlabel,"x"],[ylabel,"y"],[style,[lines,2]],
        [legend,"E0", "E1", "E2"])$
```

and we get the same plot as above.

### 2.2.2 Numerical Energies and Wave Functions using R

We use our homemade myrk 4 routine for the Runge-Kutta integration. When the file FW.R is loaded, a number of global parameters are defined. The top the the file FW.R has the lines:

```
## FW.R uses Runge-Kutta for finite well.
    dimensionless units
    v = 1 for x < 0 and x > 1
    V = 0 for 0< x < 1
    \mp@subsup{y}{}{\prime\prime}}(x)+gam2*(E-V(x))*y(x)=
    gam2 = gam^2 = 2500
    gam = 50 = sqrt (2*m*L^2*VO/ hbar^2)
    initial global parameters:
    N = 100
    h = 0.01
    gam = 50
    gam2 = gam^2
    xdecay = 0.5 ## start yL1 integration at x = -xdecay
    ## start yR integration at x = 1 + xdecay
    ypleft = 1e-8
    ypright = 1e-8
    debug = FALSE
    wfdebug = FALSE
    cat (" gam = ",gam, " gam2 = ", gam2,"\n")
    cat (" h = ", h, ", xdecay = ", xdecay, ",ypleft = ", ypleft," ypright = ",ypright,"\n")
```

We integrate from a point $\mathbf{x}=-\mathbf{x d e c a y}$ chosen so that we can assume $\mathbf{y}(-\mathbf{x d e c a y})=0$ to the point $\mathbf{x}=0$, thus defining a grid vector $\mathbf{x L 1}$ of integration points, a vector $\mathbf{y} \mathbf{L 1}$ of values of $y(x)$ at these grid points, and a vector ypL1 of values of $y^{\prime}(x)$ at these grid points where $V=1$.

We assume an arbitrary small value ypleft for the first derivative $y^{\prime}$ at this starting point. The resulting wave function, the solution of a homogeneous equation, can be later normalized, which will, in effect, amount to choosing the correct
initial first derivative at the left starting point.
The final value of $y$ and $y^{\prime}$ thus generated become the initial values of $y$ and $y^{\prime}$ for integration through the region where $V=0,0 \leq x \leq 1$, thus generating a grid vector $\mathbf{x L 2}$ of integration points, a vector $\mathbf{y} \mathbf{L} 2$ of values of $y(x)$ at these grid points, and a vector ypL2 of values of $y^{\prime}(x)$ at these grid points where $V=0$.

The integration in the region $x>1$ is done by starting at a location $\mathbf{x}=1+\mathbf{x d e c a y}$ where we can assume $y=0$ and we again assign an arbitrary (but negative) first derivative -ypright. We then integrate toward smaller values of $x$ until we reach $x=1$. Since we are hence integrating in the direction in which the physical solution is growing, we avoid integration instability problems produced by small roundoff and integration algorithm errors.

We then multiply the vectors $\mathbf{y} \mathbf{L 1}$, $\mathbf{y p L 1}, \mathbf{y L 2}$, and $\mathbf{y p L 2}$ by a factor which assures us that the final value of $y(x)$ produced by the independent rightward and leftward integrations agree at the matching point $x=1$. The value of $y(x)$ can be made to agree at the matching point for any energy $E$. However, the resulting wave function values will still be discontinuous because the first derivatives will not agree at the matching point.

The crucial step, then, is to design a function $F(E)$, say, that is zero (to within numerical errors) when the derivatives agree at the matching point. We can then look for the locations of sign changes in $F(E)$ to locate the energy eigenvalues.

The first step needed, in order to be able to design such a function $\mathbf{F}(\mathbf{E})$, is to design a function $\mathbf{w f}(\mathbb{E})$ which uses Runge Kutta methods to find a un-normalized wave function corresponding to a given total energy $\mathbf{E}$. Here is our code for such a wave function integrator, as listed in FW.R.

```
## wf(E) creates ** un-normalized ** numerical wave functions
## using Runge-Kutta routine myrk4.
## The wave functions are stored in global vectors
##
##
##
##
## the global xR grid extends from 1 to 1 + xdecay
wf = function (E) {
    if ( (E<0) | (E > 1) ) {
        cat (" need 0 < E < 1 \n")
        return(false) }
    ncenter = N
    if ( round(ncenter) != ncenter) {
        cat (" ncenter = ",ncenter," is not an integer \n")
        return(false) }
    nleft = round(xdecay/h)
    nright = nleft
    if (wfdebug) cat (" nleft = ",nleft," ncenter = ",ncenter," nright = ",nright, "\n")
    glr = gam2*(E - 1) ## g(x) for x < 0 and x > 1
    gc = gam2*E ## g(x) for 0<x< < < 
    if (wfdebug) cat (" glr = ",glr," gc = ", gc, "\n")
    ## construct xll, yll, and ypll for -xdecay < x < 0
    derivs.decay = function (x,y) { c (y[2], - glr*y[1]) }
    xl1 = seq (- xdecay, 0, h)
    outL = myrk4( c(0,ypleft), xl1, derivs.decay)
    yl1 = outL[[1]]
    ypl1 = outL[[2]]
    ## construct xl2, yl2, and ypl2 for 0 < x < 1
    derivs01 = function (x,y) { c (y[2], - gc*y[1]) }
    xl2 = seq(0, 1, h)
    outL = myrk4( c(last(yl1), last(ypl1)), xl2, derivs01)
    yl2 = outL[[1]]
    ypl2 = outL[[2]]
```

```
## construct xr, yr, and ypr for 1 < x < 1 + xdecay
xr = seq( 1 + xdecay, 1, -h)
outL = myrk4( c(0, -ypright), xr, derivs.decay)
yr = outL[[1]]
ypr = outL[[2]]
xr = rev(xr)
yr = rev(yr)
ypr = rev(ypr)
if (wfdebug) cat (" yr (1) = ", yr[1], "\n" )
fac = yr[1] / last(yl2)
if (wfdebug) cat (" fac = ",fac, "\n")
yl1 = fac*yll
ypl1 = fac*ypl1
yl2 = fac*yl2
ypl2 = fac*ypl2
## make global xL1,xL2,xR,yL1,yL2,yR,yPL1,yPL2,yPR
xL1 <<- xl1
xL2 <<- xl2
xR <<- xr
yL1 <<- yl1
yL2 <<- Yl2
yR <<- yr
ypL1 <<- ypl1
ypL2 <<- ypl2
ypR <<- ypr }
```

The second step needed to design $\mathbf{F}(\mathbf{E})$ is to create a function dy_diff() which returns a normalized difference of the first derivatives $y_{L}^{\prime}(x)-y_{R}^{\prime}(x)$ evaluated at the matching point $x=1$. We return this difference divided by the value of $y(x=1)$.

```
## dy_diff() uses global ypL2, ypR, and yR,
## returns a normalized difference of derivatives
        (yL'(1) - yR'(1)/ yR(1)
dy_diff = function () {
    dy_left = last(ypL2)
    dy_right = ypR[1]
    (dy_left - dy_right)/abs( yR[1] ) }
```

For example,

```
> wf(0.5)
> dy_diff()
[1] 35.0717
> last(ypL2)
[1] -0.00190471
> yPR[1]
[1] -0.237432
> yR[1]
[1] 0.00671558
```

Here is our code for $\mathbf{F}(\mathbf{E})$ :

```
## energy eigenvalue if global function F(E) = 0.
## F(E) calls wf(E) then returns dy_diff(), but
##
F = function (E) {
    if (E<0 | E > 1) {
        cat (" in F(E), E = ",E," should be between 0 and 1 \n ")
        return(FALSE) }
    wf(E)
    dy_diff() }
```

Here is an example of using $\boldsymbol{F}(\mathrm{E})$ to produce a rough graphical survey of the possibilities:

```
>EL = seq(0.1,0.9,0.01)
> FL = sapply(EL, F)
> fll(FL)
    82.3613 17.2303 81
> plot(EL, FL, type = "l", lwd = 2, col = "blue", xlab = "E",ylab = "F(E)" )
> mygrid()
```

which produces the plot


Figure 18: Crude Plot of $\mathrm{F}(\mathrm{E})$
We can use $\mathbf{F}(\mathbf{E}$ ) to search for energy eigenvalue candidates. We do this with a function bracket (func, $\mathbf{x}, \mathbf{d x}, \mathbf{x a c c}$ ) which attempts to return two values of $\mathbf{x}$ at which func has the opposite sign.

```
## bracket is a modified version of bracket_basic, designed to work with
## the functions F(E) or F1(k) which can return FALSE.
## bracket looks for a sign change in func,
## starting with xx, and increasing xx by dxx each step.
## If sign change is found, then we back up to the previous xx
## and search with new dxx value one half of the previous value.
## normally returns [ea,eb], but if can't find change in sign,
## then returns [0,0], and if func returns FALSE, then
## bracket returns FALSE.
bracket = function (func, xx,dxx, xacc) {
    x = xx
    dx = dxx
    it = 0
    itmax = 1000
    anerror = FALSE
    anerror2 = FALSE
    repeat {
        it = it + 1
        if (it > itmax) {
        cat (" can't find change in sign \n")
```

```
        anerror = TRUE
        break}
        x1 = x
        x2 = x + dx
        if ( debug ) cat (" it = ",it," x1 = ", x1," x2 = ", x2," dx = ", dx, "\n")
        f1 = func(x1)
        if ( f1 == FALSE) {
        cat (" in bracket, f1 = FALSE , x1 = ", x1, " dx = ", dx, " \n ")
        anerror2 = TRUE
        break }
        f2 = func(x2)
        if ( f2 == FALSE) {
        cat (" in bracket, f2 = FALSE , x2 = ", x2, " dx = ", dx, " \n ")
        anerror2 = TRUE
        break }
    if ( f1 * f2 < 0 ) {
    if ( abs(dx) < xacc ) break
    x = x - dx
    dx = dx/2 } else x = x2 }
if (anerror) c(0,0) else if (anerror2) FALSE else c(x1,x2) }
```

Here is an example of using bracket with the function $\boldsymbol{F}(\mathbf{E})$. This example produces the zero node ground state case, and we plot the un-normalized wave function pieces produced by $\mathbf{w f}(\mathrm{E})$.

```
> out = bracket (F,0.0005,0.0001,0.00005)
> out
[1] 0.003625 0.003650
> e = uniroot(F,out, tol = 1e-16) $root
>e
[1] 0.00364983
> wf(e)
> plot(0, type = "n", xlim = c(min(xL1), max(xR)), ylim = c(0, max(yL2)),
    xlab = "x", ylab = "y" )
> lines (xL1, yL1, lwd = 3, col = "blue")
> lines (xL2, yL2, lwd = 3, col = "red")
> lines (xR, yR, lwd = 3, col = "green")
> mygrid()
```

which produces a plot of the un-normalized ground state wave function with zero nodes.


Figure 19: Numerical Un-normalized Ground State Wave Function
We can check the normalized difference in slopes at the matching point for the solution produced by $\mathbf{w f}(\mathbf{E})$ :

```
> dy_diff()
[1] -3.42736e-12
```

We can check the number of nodes with the function num_nodes ().

```
num_nodes = function () {
    n = 0
    for ( j in 1 : (length(yL2) - 1) ) { if (yL2[ j ] * yL2[ j + 1 ] < 0) n = n + 1 }
    n }
```

For the numerical ground state solution generated above by $\mathbf{w f}(\mathbf{E})$ we get:

```
> num_nodes()
[1] 0
```

A function normalize () uses the current wave function pieces produced by $\mathbf{w f}(\mathbf{E})$ and uses our Simpson's rule function simp to produce global normalized wave function (vectors) $\mathbf{x n}$ and $\mathbf{y n}$. We use the function rest defined in $\mathbf{c p} 3 . \mathrm{R}$ in normalize().

Before we use normalize (), let's show interactively the route we follow in the beginning of normalize ():

```
> xn = c (xL1, c ( rest (xL2), rest (xR) ) )
> fll(xn)
    -0.5 1.5 201
> yn =c (yL1, c (rest (yL2), rest (yR) ) )
> fll(yn)
    0 0 201
> head(xn)
[1] -0.50 -0.49 -0.48 -0.47 -0.46 -0.45
> head(yn)
[1] 0.00000e+00 1.04151e-10 2.34784e-10 4.25090e-10 7.23426e-10 1.20560e-09
> simp(xn,yn^2)
[1] 6652.68
```

Here we call normalize (), and then check the normalization interactively using Simpson's rule simp.

```
> normalize()
    AA = 6652.68
    x_mean = 0.5
    delx = 0.18802
> simp(xn,yn^2)
[1] 1
```

Here is the code for normalize ().

```
## normalize() uses the current global xL1,yL1, xL2,yL2, xR, yR and
## the utility functions rest and simp to define global
## xn and yn, with the latter being normalized.
normalize = function () {
    xn = c (xL1, c ( rest (xL2), rest (xR) ) )
    yn = c (yL1, c (rest (yL2), rest (yR) ) )
    ## we need xn to have odd number of elements to use simp
    if ( is.even ( length (xn) ) ) {
        xn = rest (xn)
        yn = rest (yn) }
    AA = simp (xn,yn^2)
    cat ( " AA = ",AA, "\n" )
yn = yn/sqrt (AA)
x_mean = simp (xn, xn * yn^2)
cat (" x_mean = ", x_mean, "\n" )
x2_mean = simp (xn, xn^2 * yn^2)
delx2 = x2_mean - x_mean^2 ## this should be positive!
delx = sqrt (delx2)
cat (" delx = ", delx, "\n" )
xn <<- xn
yn <<- yn }
```

After using normalize (), one can plot the normlized wave function (the current vectors $\mathbf{x n}$ and $\mathbf{y n}$ ) using the function yn_plot_current():

```
> yn_plot_current()
ymax = 1.3867
```

which produces the plot


Figure 20: Numerical Normalized Ground State Wave Function
The function yn_plot_current () uses the current global vectors $x n$ and $y n$ created by normalize ().

```
## yn_plot_current() uses the currently defined normalized set (xn,yn)
yn_plot_current = function () {
    ymn = floor( min(yn) )
    ymx = 1 + floor( max (yn) )
    cat (" ymax = ", max(yn), "\n" )
    plot(xn, yn, type = "l", lwd = 3, col = "blue", ylim = c(ymn, ymx),
        xlab = "x", ylab = "y", tck = 1) }
```

The more versatile function $\mathbf{y n} \_$plot $(\mathbf{E}, \mathbf{x m i n}, \mathbf{x m a x})$ does three tasks in succession, first calling $\mathbf{w f}(E)$ to create the un-normalized wave function pieces, then calling normalize () to create the normalized wave function vectors in the form of $\mathbf{x n}$ and $\mathbf{y n}$, and finally making a plot of the normalized wave function, using $\mathbf{x m i n}$ and $\mathbf{x m a x}$ to control the display.

Here is an example dealing with the first excited (one node) state.

```
> out = bracket (F,0.01,0.005,0.001)
> out
[1] 0.014375 0.015000
> e = uniroot(F,out, tol = 1e-16) $root
>e
[1] 0.0145973
> yn_plot(e,-0.5,1.6)
    E = 0.0145973
    number of nodes = 1, dy_diff = 1.61333e-13
AA = 1278.41
x_mean = 0.5
delx = 0.276562
normalized ymax = 1.38655
```

which produces the plot


Figure 21: Numerical Normalized First Excited State Wave Function
Here is our code for yn_plot:

```
## yn_plot(E,xmin, xmax) first calls wf(E) to create
## un-normalized wave functions corresponding to the
## given energy E. Then normalizes those wave functions
## to produce the vectors xn and yn. Finally makes a plot
## of yn over only the region (xmn, xmx)
yn_plot = function (E,xmn,xmx) {
    wf(E)
    cat (" E = ",E, "\n" )
    cat (" number of nodes = ", num_nodes(),", dy_diff = ",dy_diff(), "\n" )
    normalize()
    cat (" normalized ymax = ", max (yn), "\n" )
    ymn = floor( min(yn) )
    ymx = 1 + floor( max (yn) )
    plot(xn, yn, type = "l", lwd = 3, col = "blue", xlim = c(xmn,xmx), ylim = c(ymn, ymx),
        xlab = "x", ylab = "y", tck = 1) }
```

We now want to construct a function levels (. . .) which will produce a vector of the energy levels, found using our numerical Runge-Kutta methods, starting with the ground state energy, and continuing up to some maximum energy. Some experimentation shows that instead of using $\boldsymbol{F}(\mathbf{E})$ with a succession of small values of $\mathbf{E}$, it is easier to use a function $\mathbf{F 1}(\mathbf{k})$, since the $\mathbf{k}$ values corresponding to the energy eigenvalues are larger numbers of order $O(1)$. Here is such a function, which we call F1 (k).

```
## F1(k): energy eigenvalue if global function F1(k) = 0.
    F1(k) calls wf(E) then returns dy_diff(), but
        returns false if k <= 0 or k >= gam.
= function (k) {
    if (k <= 0 | k >= gam) {
        cat (" in F1(k), k = ",k," k should be greater than 0 and less than ",gam, "\n")
        return(FALSE) }
    wf(k^2/gam2)
    dy_diff() }
```

Here is an example of use of $\mathbf{F 1}(\mathbf{k})$ to find the ground state energy:

```
> F1(2.9)
[1] 34.295
> F1(3.05)
[1] -49.8828
> out = bracket (F1,2.9,0.05,0.02)
> out
[1] 3.0125 3.0250
> k0 = uniroot(F1, out, tol = 1e-16) $root
> k0
[1] 3.02069
> EO = ktoE (k0)
> EO
[1] 0.00364983
> wf(EO)
> dy_diff()
[1] 3.41072e-13
> num_nodes()
[1] 0
```

which reveals a zero node wave function with a very small value of dy_diff(), a signal of a good wave function.
We can make a crude plot of $\mathbf{F 1}(\mathbf{k})$ versus $\mathbf{k}$

```
> kL = seq (1, 49,by=0.5)
> head(kL)
[1] 1.0 1.5 2.0 2.5 3.0 3.5
> F1L = sapply (kL, F1)
> head (F1L)
[1] 50.6042 50.0387 48.9460 46.2163 13.2072 57.5400
> plot(kL, F1L, type = "l", lwd = 2,col = "blue",xlab = "k",ylab = "F1(k)")
> mygrid()
```

which produces the plot


Figure 22: Crude Plot of F1(k) versus k

Let's use $\mathbf{F 1}(\mathrm{k})$ to look at the region $2.8<=\mathrm{k}<=3.3$ :

```
> kL = seq (2.8, 3.3, by = 0.05)
> head(kL)
[1] 2.80 2.85 2.90 2.95 3.00 3.05
> F1L = sapply( kL, F1)
> head(F1L)
[1] 40.3843 37.9920 34.2950 27.7891 13.2072 -49.8828
> plot(kL, F1L, type = "l", lwd = 2,col = "blue",xlab = "k",ylab = "F1(k)")
> mygrid()
```

which produces the plot


Figure 23: F1(k) Zoom
This zoom plot shows roots at roughly $\mathbf{k}=3.01$ and $\mathbf{k}=3.06$. The first root is close to the valid ground state case as shown above. We can use bracket with $\mathbf{F 1}(\mathrm{k})$ to refine the second root.

```
> out = bracket (F1, 3.03,0.01,0.005)
> out
[1] 3.0775 3.0800
> kv = uniroot(F1, out, tol = 1e-16) $root
> kv
[1] 3.07995
> Ev = ktoE(kv)
> Ev
[1] 0.00379445
> wf(Ev)
> num_nodes()
[1] 0
> dy_diff()
[1] -2.64829e+15
```

The large value of dy_diff() shows that this second, slightly larger root, is an un-physical solution. Since we already have a valid zero node solution at the lower energy, there cannot be a second zero node solution at another, higher, energy. This pattern persists, with the physical root being smaller, and the un-physical root being slightly larger. This pattern provides the rationale for our code for levels (...). Once we have found a solution with a given number of nodes, we reject all solutions with higher energy but the same number of nodes. The unphysical roots correspond to a sudden change in which (left or right) integration function has the larger slope magnitude at the matching point.

Here is an example of using levels.

```
> EL = levels(1, 8,0.05,0.02)
-------------
kstart = 1 dk = 0.05
ka = 3.0125 kb = 3.025
kroot = 3.02069
eroot = 0.00364983
number of nodes = 0
dy_diff at x = 1 is 3.41072e-13
input c or s
c
---------------- levels
nlast = 0
kstart = 3.125 dk = 0.05
ka = 6.0375 kb = 6.05
kroot = 6.04096
eroot = 0.0145973
number of nodes = 1
dy_diff at x = 1 is 1.61333e-13
input c or s
c
---------------- levels
nlast = 1
kstart = 6.2 dk = 0.05
ka = 9.05 kb = 9.0625
kroot = 9.06036
eroot = 0.032836
number of nodes = 2
dy_diff at x = 1 is -1.36136e-13
input c or s
c
> EL
[1] 0.00364983 0.01459726 0.03283602
```

We can then use $\mathbf{y n} \_$plot $(\mathbf{E}, \mathbf{x m i n}, \mathbf{x m a x})$ to both construct the vectors $\mathbf{x n}$ and $\mathbf{y n}$ of the normalized wave function and make a plot. We save the normalized wave functions by assignment statements such as $\mathbf{x n 0}=\mathbf{x n}$, and $\mathbf{y n 0}=\mathbf{y n}$, before another call to $\mathbf{y n}$ _plot defines the wave functions corresponding to a different energy. In the following, we do not show the plots produced by the calls to yn_plot.

```
> yn_plot(EL[1],-0.5,1.5)
E = 0.00364983
number of nodes = 0, dy_diff = 3.41072e-13
AA = 6652.68
x_mean = 0.5
delx = 0.18802
normalized ymax = 1.3867
> xnO = xn
>fll(xn0)
    -0.5 1.5 201
> ynO = yn
> fll(yn)
    0 0 201
```

We continue in this manner with the first excited state and the second excited state.

```
> yn_plot(EL[2],-0.5,1.5)
E = 0.0145973
    number of nodes = 1 , dy_diff = 1.61333e-13
AA = 1278.41
x_mean = 0.5
delx = 0.276562
normalized ymax = 1.38655
> xn1 = xn
> fll(xn1)
    -0.5 1.5 201
> yn1 = yn
> fll(yn1)
    0 0 201
```

```
> yn_plot(EL[3],-0.5,1.5)
E = 0.032836
number of nodes = 2, dy_diff = -1.36136e-13
AA = 365.384
x_mean = 0.5
delx = 0.290108
normalized ymax = 1.38569
> xn2 = xn
> yn2 = yn
```

We can then combine the plots for the wave functions of these three lowest lying states.

```
> plot(0,type = "n",xlim = c(-0.5,1.5),ylim = c(-1.5,1.5),xlab="x",ylab="y")
> lines(xn0,yn0,lwd=2,col="blue")
> lines (xn1,yn1,lwd=2,col="red")
> lines(xn2,yn2,lwd=2,col="green")
> mygrid()
> legend("bottomright",col = c("blue","red","green"),
+ legend = c("EO", "E1", "E2"), lwd=2,cex=1.5)
```

which produces the plot


Figure 24: Finite Well: Lowest Three Energy Level Wave Functions
Just for practice, we can write these three wave functions to a file in the local folder, and then restart $\mathbf{R}$, read in the file contents, and make a plot based on the file contents.

The simplest approach is to use save (filePath,a1,a2,a3, ...), where objects a1, a2, etc are object names bound to quantities known to R . The names and the objects bound to the names are stored in a binary file format in the file requested (which is created if it does not yet exist, and overwritten if it already exists).

One can use load to load in that file into a new session, and the names and objects will then be available for use in your new Maxima session. In the following, we first save the wave function files to xy.rda. We then use rm to remove knowledge of those objects from the current session. We then use load to recover knowledge of those objects, which can then be used as before, for example to make plots and make calculations.

```
> save(xn0,yn0,xn1,yn1,xn2,yn2, file = "xy.rda")
> rm(xn0,yn0,xn1,yn1,xn2,yn2)
```

```
> fll(xn0)
Error in fll(xn0) : object 'xn0' not found
> load("xy.rda")
> fll(xn0)
    -0.5 1.5 201
```

For example, we can remake the plot of all three wave functions

```
> plot(0,type = "n",xlim = c(-0.5,1.5),ylim = c(-1.5,1.5),xlab="x",ylab="y")
> lines(xn0,yn0,lwd=2,col="blue")
> lines (xn1,yn1, lwd=2,col="red")
> lines(xn2,yn2,lwd=2,col="green")
> mygrid()
> legend("bottomright",col = c("blue","red","green"),
+ legend = c("EO", "E1", "E2"), lwd=2,cex=1.5)
```

and we get the same plot as above.
Here is the code for levels:

```
## levels(kmin,kmax, dk, kacc ) returns a vector c( Ea, Eb,...) of energy levels with
    increasingly larger number of nodes in energy range (Emin, Emax)
    according to Emin = kmin^2/gam^2, and Emax = kmax^2/gam^2.
    uses F1(k) (inside bracket) to find roots, and calls wf(E), num_nodes() and
        dy_diff() for each root found,
    Uses bracket and uniroot.
    The arguments (dk, kacc) are used to call bracket, and do not describe
        the accuracy of the energy levels found.
    Once a good energy e.v. is found we look for the region of
        energies with one more node and search there.
    Code has interactive continue or stop.
    Searching for the k eigenvalues via F1(k) is easier than searching
        directly for the E eigenvalues via F(E) for the case gam = 50 we
        consider in this example.
levels = function (kmin,kmax,dk, kacc ) {
    rmax = 20
    eL = rep(NA, rmax) ## vector eL will hold energy eigenvalues found
k = kmin
nlast = -1
j = 1
repeat {
    if (k > kmax | j > rmax) break ## exit do loop
    cat ("---------------- levels ------------------------------
    cat (" nlast = ", nlast,"\n")
    cat (" kstart = ", k," dk = ", dk, "\n" )
    out = bracket(F1,k, dk, kacc)
    cat (" ka = ",out[1]," kb = ",out[2],"\n")
    if (out[1] == 0) {
        cat (" can't find bracket interval \n")
            cat (" k = ",k, "\n")
            break }
    kroot = uniroot(F1, out, tol = 1e-16) $root
    cat (" kroot = ", kroot, "\n")
    eroot = kroot^2/gam2
    cat (" eroot = ", eroot, "\n")
    wf(eroot)
    nn = num_nodes()
    cat (" number of nodes = ", nn, "\n")
    cat (" dy_diff at x = 1 is ", dy_diff(), "\n" )
    eL[ j ] = eroot
    nlast = nn
    j = j + 1
```

```
    r = readline (" input c or s \n ")
    if (r == "s") break ## exit do loop
    ## search for a k value greater than kb which produces
    ## a wave function with nn + 1 nodes
    knext = out[2] + dk
    repeat {
        wf(knext^2/gam2)
        if (num_nodes() > nlast) {
            k = knext
            break } else knext = knext + dk } } ## end of outer repeat loop
## remove NA's at end of eL
eL[!is.na(eL)] }
```


## 3 The Numerov Integration Method

Numerov's method was developed by the Russian astronomer Boris Vasil'evich Numerov in the years 1924-1927. Numerov's algorithm is a simple and efficient method for integrating linear second order ode's which do not contain a first order derivative term and is especially useful for homogeneous ode's, such as Schroedinger's equation.

Corresponding to a grid of equally spaced values $x_{n}$ of the independent variable $x$, will be values $y_{n}$ of the dependent variable. A numerical solution of the ode

$$
\begin{equation*}
y^{\prime \prime}(x)+g(x) y(x)=S(x) \tag{3.1}
\end{equation*}
$$

can then be constructed using the following Numerov three term recursion relation

$$
\begin{equation*}
\left(1+\frac{h^{2}}{12} g_{n+1}\right) y_{n+1}-2\left(1-\frac{5 h^{2}}{12} g_{n}\right) y_{n}+\left(1+\frac{h^{2}}{12} g_{n-1}\right) y_{n-1}=\frac{h^{2}}{12}\left(S_{n+1}+10 S_{n}+S_{n-1}\right)+O\left(h^{6}\right) \tag{3.2}
\end{equation*}
$$

Solving this linear equation for either $y_{n+1}$ or $y_{n-1}$ then provides a recursion relation for integrating either forward or backward in $x$, with a local error $O\left(h^{6}\right)$. The Numerov scheme is more efficient than the Runge-Kutta method, as each step of the Numerov method requires the computation of $g$ and $S$ at only the grid points (and not at intermediate points). However, the Runge-Kutta method provides both $y(x)$ and $y^{\prime}(x)$ at the grid points, whereas the Numerov method only provides $y(x)$ at the grid points.

For a derivation of the Numerov method, see
https://en.wikipedia.org/wiki/Numerov's_method

### 3.1 Classical Simple Harmonic Oscillator Test Case

Let's try out the Numerov method for a classical simple harmonic oscillator with unit period, defined by

$$
\begin{equation*}
\frac{d^{2} y}{d x^{2}}=-4 \pi^{2} y(x), \quad y(0)=1, \quad y^{\prime}(0)=0 \tag{3.3}
\end{equation*}
$$

over the domain $0 \leq x \leq 1$. This corresponds to (3.1) with $S(x)=0$ and $g(x)=4 \pi^{2}$, in which case the Numerov algorithm (3.2) takes the form (integrating in the direction of increasing $x$ ):

$$
\begin{equation*}
y_{n+1}=A y_{n}-y_{n-1} \tag{3.4}
\end{equation*}
$$

with

$$
\begin{equation*}
A=\frac{2\left(1-\frac{5 h^{2} \pi^{2}}{3}\right)}{\left(1+\frac{h^{2} \pi^{2}}{3}\right)} \tag{3.5}
\end{equation*}
$$

The analytic solution for the initial conditions assumed in (3.3) is $y(x)=\cos (2 \pi x)$. Given $y_{0}=y(x=0)$ and $y_{0}^{\prime}=y^{\prime}(x=0)$ we can calculate $y_{1}=y(h)$ using a Taylor series expansion about $x=0$. In order to calculate $y_{2}=y(2 h)$ with an accuracy $O\left(h^{6}\right)$ we need $y_{1}$ with this same accuracy. It is sufficient, however, to calculate $y_{1}$ with an accuracy $O\left(h^{5}\right)$ because the global error of Numerov's method is $O\left(h^{5}\right)$ and we calculate $y_{1}$ just once. Using the expansion

$$
\begin{equation*}
y_{1}=y(h)=y_{0}+h y_{0}^{\prime}+\frac{h^{2}}{2} y^{\prime \prime}(0)+\frac{h^{3}}{3!} y^{\prime \prime \prime}(0)+\frac{h^{4}}{4!} y^{\prime \prime \prime \prime}(0)+O\left(h^{5}\right) \tag{3.6}
\end{equation*}
$$

we get

$$
\begin{equation*}
y_{1}=y_{0}+h y_{0}^{\prime}-2 \pi^{2} h^{2} y_{0}-\frac{2}{3} \pi^{2} h^{3} y_{0}^{\prime}+\frac{2}{3} \pi^{4} h^{4} y_{0} \tag{3.7}
\end{equation*}
$$

### 3.1.1 Classical SHO Numerov Method Using Maxima

We have written two Maxima "do loop" versions of Numerov's method to integrate our classical simple harmonic oscillator example. These two versions are called sho ( $\mathrm{h}, \mathrm{y} 0, \mathrm{yp} 0$ ) and $\mathbf{s h o} 2(\mathrm{~h}, \mathrm{y} 0, \mathrm{yp} 0$ ) and they are in the file cp3.mac. Here is the first version, which builds up a list of lists $\boldsymbol{r L}$ using the Maxima function cons.

```
/* sho(h,y0,yp0)
integrates simple harmonic oscillator with unit period
    d^2y/dx^2 = - 4 pi^2 y(x) over [x,0,1]
    using the numerov method.
    Input: h = step size, y0 = y(0), yp0 = y' (0)
    Output: list [[0,y0],[h,y1],[2 h, y2],...]
*/
sho(h,y0,yp0) :=
block ([A, Y1,N,Ym,Yz,YP,rL, x,
            xmin:0, xmax:1, numer], numer:true,
    A : 2* (1 - 5*%pi^2*h^2/3)/(1 + %pi^2*h^2/3),
    y1 : y0 + h*yp0 - 2*%pi^2*h^2*y0 - 2*%pi^2*h^ 3*yp0/3 +
                2*%pi^4*h^4*y0/3,
    N : round( (xmax - xmin)/h),
    rL : [[xmin+h,y1], [xmin,y0]],
    x : xmin + 2*h,
    ym : y0,
    yz : y1,
    for j thru N - 1 do (
        yp : A*yz - ym,
        rL : cons ( [x,yp], rL),
        ym : yz,
        yz : yp,
        x : x + h),
    reverse(rL)) $
```

This code uses $\mathbf{y p}$ ("y plus") to represent $y_{n+1}, \mathbf{y z}$ (" $\mathbf{y}$ zero") to represent $y_{n}$, and $\mathbf{y m}$ ("y minus") to represent $y_{n-1}$, hence the line $\mathbf{y p}: \mathbf{A} * \mathbf{y z}$ - $\mathbf{y m}$, in the loop, with values being "rolled" at the end of the loop. If we use $h=0.01$, we get good agreement with the exact solution:

```
(%i1) load(cp3);
(%o1) "c:/k3/cp3.mac"
(%i2) soln : sho(0.01, 1, 0)$
(%i3) xL : take(soln,1)$
(%i4) fll(xL);
(%04) [0,1.0,101]
(%i5) head(xL);
(%05) [0,0.01,0.02,0.03,0.04,0.05]
(%i6) tail(xL);
(%06) [0.95,0.96,0.97,0.98,0.99,1.0]
(%i7) xL[1];
(%07) 0
(%i8) yL : take(soln,2)$
```

```
(%i9) fll(yL);
(%09) [1,1.0,101]
(%i10) head(yL);
(%010) [1,0.998027,0.992115,0.982287,0.968583,0.951057]
(%i11) yL[1];
(%011) 1
(%i12) plot2d([['discrete,xL, yL], cos(2*%pi*x)],['x,0,1],['xlabel,"x"],
    ['ylabel,""],['legend, "numerov","analytic"],
        [gnuplot_preamble,"set key bottom"])$
(%i13) plot2d([['discrete,soln], cos(2*%pi*x)],['x,0,1],['xlabel,"x"],
    ['ylabel,""],['legend, "numerov","analytic"],
        [gnuplot_preamble,"set key bottom"])$
```

Either of the above plot2d statements produce the plot


Figure 25: Classical SHO Numerov Solution with $h=0.01$
The second Maxima version sho2 uses 'hashed arrays' called $\mathbf{x L}$ and $\mathbf{y}^{\mathrm{L}}$ in the code. By using a line in the code of the form local ( $\mathbf{x L}, \mathrm{YL}$ ) these hashed arrays are not known at the global level, and we can reuse the same names globally as shown in the example below. Note that the form of the output of sho 2 is different than that of sho.

```
sho2(h, y0, yp0) :=
block([A,y1,xmin, xmax,N,x,ym,yz,yp, numer], numer:true,
    local(xL, yL),
    A : 2*(1 - 5*%pi^2*h^2/3)/(1 + %pi^2*h^2/3),
    y1 : y0 + h*yp0 - 2*%pi^2*h`^2*y0 - 2*%pi^2*h^3*yp0/3 +
        2*%pi^4*h^4*y0/3,
    print(" A = ",A," y1 = ", y1),
    xmin : 0,
    xmax : 1,
    N : round ( (xmax - xmin)/h),
    print(" N = ", N),
    xL[1] : xmin,
    xL[2] : xmin + h,
    yL[1] : y0,
    yL[2] : y1,
    x : xmin + 2*h,
    ym : y0,
    yz : y1,
    for j : 3 thru N + 1 do (
        yP : A*yz - ym,
        xL[j] : x,
        YL[j] : Yp,
        ym : yz,
```

```
    yz : yp,
    x : x + h ),
[ listarray(xL), listarray(yL) ] )$
```

We use the same input parameters as before in this example. To emphasize the fact that the code line local ( $\mathbf{x L}, \mathrm{y} \mathbf{L}$ ) hides the values assigned to the hashed arrays, we first use remvalue to remove the definition of the lists $\mathbf{x L}$ and $\mathbf{y}^{\boldsymbol{L}}$ produced in the example above.

```
(%i14) remvalue(xL, yL);
(%O14) [xL,yL]
(%i15) xL[1];
(%O15) xL[1]
(%i16) yL[1];
(%o16) yL[1]
(%i17) soln : sho2(0.01, 1, 0) $
A = 1.9960535 y1 = 0.998027
N = 100
(%i18) xL[1];
(%O18) xL[1]
(%i19) xL : soln[1]$
(%i20) fll(xL);
(%O20) [0,1.0,101]
(%i21) head(xL);
(%021) [0,0.01,0.02,0.03,0.04,0.05]
(%i22) yL[1];
(%O22) yL[1]
(%i23) yL : soln[2]$
(%i24) fll(yL);
(%O24) [1,1.0,101]
(%i25) plot2d([['discrete,xL,yL], cos(2*%pi*x)],['x,0,1],['xlabel,"x"],
    ['ylabel,""],['legend, "numerov","analytic"],
        [gnuplot_preamble,"set key bottom"])$
```

which produces exactly the same plot as produced using sho.

### 3.1.2 Classical SHO Numerov Method Using R

The file $\mathrm{cp} 3 . \mathrm{R}$ contains a function sho ( $\mathrm{h}, \mathrm{y} 0, \mathrm{yp} 0$ ) with the same input syntax as the Maxima version. But the R function returns a R list: list ( $\mathbf{x L}, \mathrm{yL}^{\mathrm{L}}$ ) in which $\mathbf{x L}$ is a vector containing the $x$ positions, and $\mathbf{y L}$ is a vector containing the corresponding $y$ values produced by the Numerov code.

```
## sho(h, y0, yp0)
    integrates simple harmonic oscillator with unit period
    d^2y/dx^2 = - 4 pi^2 y(x) over [x,0,1]
    using the numerov method.
    Input: h = step size, y0 = y(0), yp0 = y' (0)
    Output: list( xL, yL)
sho = function(h,y0,yp0) {
    A = 2*(1 - 5*pi^2*h^2/3)/(1 + pi` 2*h`^2/3)
    y1 = y0 +h*yp0 - 2*pi^2*h` 2*y0 - 2*pi^2*h^ 3*yp0/3 +
            2*pi^4*h^4*y0/3
    xmin = 0
    xmax = 1
    N = (xmax - xmin)/h
    yL = vector(length = N + 1)
    xL = vector(length = N + 1)
    xL[1] = xmin
    xL[2] = xmin +h
    yL[1] = y0
    yL[2] = y1
```

```
x = xmin + 2*h
ym}=\mp@subsup{y}{0}{0
yz = y1
for (j in 3:(N + 1)) {
    yp = A*yz - ym
    xL[j] = x
    YL[j] = YP
    ym = yz
    yz = yp
    x = x + h }
list( xL, yL) }
```

We can then compare the Numerov method with the analytic solution with the same value of $h$ as used in our Maxima work above. The R function $\mathbf{f l l}$ (also in $\mathbf{c p} \mathbf{3 . R}$ ) prints out the first, last, and length of a R vector.

```
> source("cp3.R")
> soln = sho(0.01,1,0)
A = 1.99605 y1 = 0.998027
    N = 100
> xL[1]
Error: object 'xL' not found
> xL = soln[[1]]
> fll(xL)
    O 1 101
> yL = soln[[2]]
> fll(yL)
    1 1 101
> plot(xL,yL,type="l",lwd=2,xlab="x",ylab = "")
> curve(cos(2*pi*x) , 0,1,add=TRUE,col="red",lwd=2)
> mygrid()
> legend("bottomright",col=c("black","red"),cex=1.5,
+ legend=c("Numerov","Analytic"),lwd=2)
```

which produces a plot which shows agreement:


Figure 26: Numerov Solution for Classical SHO with $h=0.01$ using R

## 4 The Lennard Jones 6-12 Potential Well: Energy Levels and Wave Functions

The Lennard Jones 6-12 potential (energy) can be defined by

$$
\begin{equation*}
V(r)=4 V_{0}\left[\left(\frac{a}{r}\right)^{12}-\left(\frac{a}{r}\right)^{6}\right] \tag{4.1}
\end{equation*}
$$

in which $a$ is an adjustable length, $V_{0}$ is an adjustable energy and $r \geq 0$. We used the quasi-classical WKB method in Example 1 of this series to estimate the lowest molecular energy levels associated with this potential.

We define a dimensionless coordinate $\tilde{x}=r / a$, a dimensionless potential energy $\tilde{V}=V / V_{0}$ and energy $\tilde{E}=E / V_{0}$ and a dimensionless wave function $\tilde{\psi}(\tilde{x})=\sqrt{a} \psi(x)$, in terms of which we have a dimensionless potential (energy)

$$
\begin{equation*}
\tilde{V}(\tilde{x})=4\left[\frac{1}{\tilde{x}^{12}}-\frac{1}{\tilde{x}^{6}}\right] \tag{4.2}
\end{equation*}
$$

and Schroedinger's equation takes the form

$$
\begin{equation*}
\frac{d^{2} \tilde{\psi}(\tilde{x})}{d \tilde{x}^{2}}+\gamma^{2}(\tilde{E}-\tilde{V}(\tilde{x})) \tilde{\psi}(\tilde{x})=0 \tag{4.3}
\end{equation*}
$$

in which the dimensionless parameter

$$
\begin{equation*}
\gamma=\left[\frac{2 m a^{2} V_{0}}{\hbar^{2}}\right]^{1 / 2} \tag{4.4}
\end{equation*}
$$

The wave function normalization condition becomes

$$
\begin{equation*}
\int_{0}^{\infty} \tilde{\psi}(\tilde{x})^{2} d \tilde{x}=1 \tag{4.5}
\end{equation*}
$$

In Example 1 we used the case $\gamma=50$, which we will also use here.
In the following, we omit the tildes and use $y(x)$ to represent $\tilde{\psi}(\tilde{x})$.

### 4.1 The Numerov Method Using Maxima

The file LJ6-12.mac contains a group of Maxima functions designed to explore the energy levels and wave functions associated with a quantum particle in the Lennard-Jones $6 / 12$ potential introduced in the previous section. The dimensionless parameter $\gamma$ takes on the same value as used for our quasi-classical limit approach in Example 1 of this series, and gam2 in our code represents $\gamma^{2}$.

The dimensionless potential (energy) $V(x)$ does not depend on the value of $\gamma$ and we can make a simple plot and explore its shape. After defining a function which is based on the form of $V(x)$, we define the value of $x$, called $\mathbf{x m}$, where $V(x)$ takes on its minimum value, and show that $V(x m)=-1, V(1)=0$, and $V(x)$ approaches large positive values as $x$ approaches 0 , and small negative values for $x$ very large. Recall that now $x$ represents a non-negative dimensionless number.

```
(%i1) V(z) := ( 4*(z^(-12) - z^(-6)) ) $
(%i2) xm : float(2^(1/6));
(%O2) 1.122462048309373
(%i3) V(xm);
(%03) -1.0
(%i4) V(1);
(%O4) 0
(%i5) V(0.5);
(%05) 16128.0
(%i6) V(3);
(%06) -2912/531441
(%i7) float (%);
```

```
(%O7) -0.0054794417442388
(%i8) V(0.1);
(%08) 3.9999959999999946E+12
(%i9) V(10.0);
(%09) -3.9999959999999997E-6
%i10) limit(V(x),x,inf);
(%010) 0
(%i11) limit( V(x),x,0,plus );
(%011) inf
(%i12) plot2d([ [discrete,[ [0.8,0],[2,0]]], V(x)],
    [x,0.8,2],[y,-1.5,2], [xlabel, "x"],[ylabel,"V(x)"],
    [style, [lines,3]],[legend,false],[gnuplot_preamble,"set grid"])$
plot2d: some values were clipped.
```

which produces the dimensionless potential (energy) plot:


Figure 27: Dimensionless Lennard Jones Potential (Energy)
The bound state energy eigenvalues lie in the range $-1<E<0$ and the classical turning points are defined by the equation $E=V(x)$. Setting $y=x^{6}$, one obtains a quadratic equation in $y$ which is easily solved for $y_{t p}$. One can then obtain the classical turning points as a function of $E$ from $x_{t p}=y_{t p}^{1 / 6}$. Bearing in mind that $E<0$, we can write the turning points in Maxima code as

```
(%i13) xin(E) := xm*(sqrt(E+1)/E-1/E)^(1/6) $
(%i14) xout(E) := xm*(sqrt(E+1)+1)^(1/6)/(-E)^(1/6)$
```

and then one can add a hypothetical energy level line to our potential energy plot:

```
(%i15) plot2d([ [discrete,[ [0.8,0],[2,0]]], V(x),
    [discrete, [ [xin(-0.5),-0.5],[xout(-0.5),-0.5] ]] ],
    [x,0.8,2],[y,-1.5,2], [xlabel,"x"], [ylabel,"V(x)"],
        [style, [lines,3]],[legend,false],[gnuplot_preamble,"set grid"])$
plot2d: some values were clipped.
```

which produces the plot


Figure 28: Adding a Hypothetical Energy Level

Loading in the file LJ6-12 .mac defines a set of global parameters and functions defined at the top of the file:

```
/* initial global parameters: */
( h : 0.01,
    h2 : h^2/12, /* Numerov constants */
    h52 : 5*h2,
    gam2 : 2500, /* square of gam = 50 */
    x1decay : 0.3, /* start yL integration at (x1 - x1decay) */
    x2decay : 1, /* start yR integration at (x2 + x2decay) */
    y2left : le-19, /* y(x_left + h) value chosen from E = -0.9 case */
    y2right : le-16, /* y(x_right - h) value chosen from E = -0.9 case */
    print(" gam2 = ", gam2),
    print(" h = ", h, ", x1decay = ", x1decay, ", x2decay = ", x2decay),
    print(" y2left = ",y2left, " y2right = ", y2right ),
    xm : float(2^(1/6)), /* this is where V(x) = -1 = minimum value */
    /* for given energy -1 < E < 0 , these are the turning points */
    xin(E) := xm*(sqrt (E+1)/E-1/E)^ (1/6),
    xout(E) := xm*(sqrt(E+1)+1)^(1/6)/(-E)^(1/6),
    /* dimensionless potential V(x) for Lennard-Jones 6/12 potential */
    V(z) := ( 4* (z^(-12) - z^(-6)) ) ) $
```

The grid size $h$ and two Numerov method constants $h 2$ and $h 52$ which depend on $h$ are then globally available. We call the left classical turning point $\mathbf{x 1}$ and we start the rightward Numerov integration at position $\mathbf{x 1}-\mathbf{x} \mathbf{1 d e c a y}$ with $y=0$. Likewise we call the right classical turning point $\mathbf{x} 2$ and start the leftward Numerov integration at position $\mathbf{x} 2+\mathbf{x} 2$ decay with $y=0$.

```
(%i1) load(cp3);
(%o1) "c:/k3/cp3.mac"
(%i2) load("LJ6-12.mac");
gam2 = 2500
h = 0.01, x1decay = 0.3, x2decay = 1
y2left = 1.0E-19 y2right = 1.0E-16
(%o2) "c:/k3/LJ6-12.mac"
```

Let us ignore, at first, some slight refinements we have in the code, and give a simplified version.

We first define a grid point $\mathbf{x} \mathbf{2 c}$ which is close to the right classical turning point $\mathbf{x} \mathbf{2}$.

We next generate a list called $\mathbf{y}^{\mathrm{L}}$ which contains the values of $y(x)$ from the grid point $\mathbf{x} \_\mathbf{l e f t}: \mathbf{x} \mathbf{1} \mathbf{-} \mathbf{x} \mathbf{1 d e c a y}$, with $y\left(x_{l e f t}\right)=0$, and $y\left(x_{l e f t}+h\right)=\mathbf{y} 2 l e f t$, and further points generated using the Numerov method, continuing to the grid point $\mathbf{x} \mathbf{2 c}+\mathbf{h}$.

Next we generate a list called $\mathbf{y R}$ which contains the values of $y(x)$ from $\mathbf{x} \mathbf{2} \mathbf{+} \mathbf{x} 2$ decay, with $y\left(x_{\text {right }}\right)=0$, and $y\left(x_{r i g h t}-h\right)=\mathbf{y}^{2} \mathbf{r i g h t}$, and further points generated using the Numerov method, continuing to the grid point $\mathbf{x} \mathbf{2 c}-\mathbf{h}$.

We then multiply all values of $y_{L}$ by a common factor which ensures that $y_{L}$ and $y_{R}$ agree on the value of $\mathbf{y}(\mathbf{x}=\mathbf{x} \mathbf{2 c})$.

The $\mathbf{x}$ grid values $\mathbf{x L}$ and $\mathbf{x R}$, and the corresponding $\mathbf{y}(\mathbf{x})$ grid values contained in the lists $\mathbf{y L}$ and $\mathbf{y R}$ are available as global quantities.

Here is code for $w f(E)$ from $\mathbf{L J 6 - 1 2}$. mac that generates the un-normalized wave functions in the form of the global lists $\mathbf{x L}, \mathbf{y} \mathbf{L}, \mathbf{x R}, \mathbf{y R}$. In the function $\mathbf{w f ( E )}$ we use the hashed arrays $\mathbf{x l}, \mathbf{y} \mathbf{l}, \mathbf{x r}$, and $\mathbf{y r}$. Their names are included in a local statement of the form local ( $\mathbf{g}, \mathbf{x l}, \mathbf{y} \mathbf{l}, \mathbf{x r}, \mathbf{y r}$ ), so that the hashed arrays are not available at the global level.
 contents of the hashed arrays available as ordinary lists globally.

```
/* wf(E) creates ** un-normalized ** wave functions
    for the Lennard-Jones 6/12 potential case.
    limits of numerical integration, x_left and x_right are
    determined by energy E and xdecay values.
    The wave functions are stored in global xL, yL, xR, yR.
    Program also defines **global** x2c, nleft, nright, x_left, x_right.
    See example run at end.
    nleft = the number of steps from x_left to x2c = grid point nearest to
    x1 = xin(E) = classical turning point < xm = 1.122462
    x2 = xout(E) = classical turning point > xm.
    the global xL grid extends from x_left to x2c + h and
    the global xR grid extends from x2c - h to x_right,
    so we can compute y'(x2c) using a 3 pt. symmetric formula.
*/
wf(E) := block( [x1,x2,x, fac,numer], numer:true,
    if (E > O) or (E < -1) then (
        print(" need -1 < E < 0 "),
        return(false)),
    local(g,xl,yl,xr,yr) ,
    g(z) := gam2*( E - V (z) ), /* coeff. func. in ode: y'r}(x)+g(x) y(x) = 0 *
    x1 : xin(E), /* classical turning point for x < xm */
    x2 : xout(E), /* classical turning point for x> xm */
    if wfdebug then print(" x1 = ", x1, " x2 = ", x2),
    x_left : x1 - x1decay,
    nleft : round ( (x2 - x_left)/h ), /* number of steps from x_left to x2c = match point */
    x2c : x_left + h*nleft,
    if wfdebug then print(" x_left = ",x_left," nleft = ",nleft," x2c = ", x2c),
    if wfdebug then print(" y2left = ", y2left, " y2right = ", y2right),
    nright : round ( (x2 + x2decay - x2c)/h ), /* number of steps from x2c to x_right */
    x_right : x2c + h*nright,
    if wfdebug then print(" nright = ",nright," x_right = ",x_right),
    /* find yL for x_left <= x <= x2c + h using Numerov algorithm */
    xl[1] : x_left,
    xl[2] : x_left + h,
    yl[1] : 0,
    yl[2] : y2left,
    for j:2 thru nleft + 1 do (
        x : x_left + j*h,
        xl[j+1] : x,
        yl[j+1] : ( 2*(1 - h52*g(x-h)) * yl[j] - (1 + h2*g(x-2*h)) * yl[j-1] )/ (1 + h2*g(x)) ),
    /* find yR for x2c - h <= x <= x_right using Numerov method */
    xr[nright + 2 ] : x_right,
    xr[nright + 1] : x_right - h,
    yr[nright + 2] : 0,
    yr[nright + 1] : y2right,
    for j:nright+1 step -1 thru 2 do (
        x : x2c + h* (j -3),
        xr[j-1] : x,
        yr[j-1] : (2*(1-h52*g(x+h))*yr[j] - (1 + h2*g(x+2*h))*yr[j+1] )/(1 + h2*g(x)) ),
    fac : yr[2]/yl[nleft + 1], /* yR(x2c) / yL(x2c) */
    for j thru nleft+2 do yl[ j ] : fac * yl[ j ],
    xL : listarray(xl),
    yL : listarray(yl),
    xR : listarray(xr),
    yR : listarray(yr),
    done) $
```

An example of the use of $\mathbf{w f}(\mathbf{E})$ for a randomly chosen energy is

```
(%i3) wf(-0.5);
(%o3) done
(%i4) fll(xL);
(%04) [0.726743,1.3867425,67]
(%i5) fll(xR);
(%05) [1.3667425, 2. 3767425, 102]
(%i6) head(yL);
(%06) [0,1.75654725E-29,-3.97814273E-28,1.06730999E-26,-3.66490209E-25,
    1.85515641E-23]
(%i7) tail(yL);
(%07) [1.46282504E-4,4.50817194E-4,7.49059762E-4,0.00104033,0.00132673,0.00161295]
(%i8) dy_diff();
(%08) 33.121756
(%i9) num_nodes();
(%09) 3
```

The function dy_diff() used above is designed to return the difference of the approximate numerical first derivatives at the matching point $\mathbf{x} \mathbf{2 c}$ implied by $\mathbf{y L}$ and $\mathbf{y R}$, divided by the value of $\mathbf{y}(\mathbf{x} \mathbf{2 c})$. Here is our code for dy_diff().

```
/* dy_diff() uses global yL, yR,nleft, h
    computes numerical y' (x2c) using
    symmetric three point method for
    both YL and YR, and returns the difference
    divided by y(x2c)
*/
dy_diff() :=
block([ypL, ypR, numer],numer:true,
    ypL : ( last (yL) - yL[nleft] ) / (2*h),
    ypR : (yR[3] - first(yR)) / (2*h),
    (ypL - ypR)/ abs (yR[2]) )$
```

The function num_nodes () has the definition:

```
/* count the number of nodes in }\mp@subsup{y}{}{L
    ignore region where elements of yL are
    tiny in magnitude.
    uses position(...) */
num_nodes() :=
block([x11, j0, yLm2, n, numer], numer:true,
    x11 : x_left + x1decay,
    j0 : position(x11, xL),
    yLm2 : rest (yL, -2), /* ignore y (x2c) and y (x2c+h) values in count */
    n : 0,
    for j : j0 thru (length(yLm2) - 1) do
        if yLm2[j] * yLm2[j + 1] < 0 then n : n + 1,
    n) $
```

and uses position(...):

```
/* position(xv, aL) is designed to be used with xL to locate
    position of first element which is equal to or greater than x1,
    since in this package xL has just increasing positive numbers in it
*/
position(xv, aL) := ( first (sublist_indices(aL,lambda([x], x >= xv))))$
```

We search for energy eigenvalues by seeking energies $\mathbf{E}$ such that the value returned by dy_diff() is zero to within numerical errors. A function $\mathbf{F}(\mathbf{E})$ allows us to scan energy ranges for energy eigenvalues.

```
/* energy eigenvalue if global function F(E) = 0 .
    F(E) calls wf(E) then returns dy_diff(), but
    returns false if E > 0.
    */
F(E) :=
block( [ numer],numer:true,
    if E > O then (
        print(" in F(E), E = ",E," should be negative "),
        return(false)),
    wf(E),
    dy_diff())$
```

Here is an example of using $\mathbf{F}(\mathbf{E})$.

```
(%i10) EL : makelist(e,e,-0.91,-0.85,0.01);
(%010) [-0.91,-0.9,-0.89,-0.88,-0.87,-0.86,-0.85]
(%i11) FL : map(F,EL);
(%011) [7.199669,2.035782,-5.228712,-24.50145,-58.60732,232.5781,61.42046]
(%i12) F(-0.91);
(%012) 7.1996689
(%i13) F(-0.88);
(%013) -24.501452
(%i14) e : find_root(F, -0.91,-0.88);
(%014) -0.896404
```

A function wf_plot (E) generates a non-normalized numerical solution using $\mathbf{w f ( E )}$, makes a plot and prints out the energy and maximum y value, the number of nodes, and the value of dy_diff() corresponding to the chosen energy $\mathbf{E}$.

```
(%i15) wf_plot(e);
E = -0.896404, ymax = 27.422334
number of nodes = 0 , dy_diff = 7.94875172E-14
```

which produces the plot


Figure 29: Zero Nodes Un-normalized Eigenfunction

The code for wf_plot (E) is

```
/*
    wf_plot (E) calls wf(E) and plot2d
    creates ** un-normalized ** wave functions
    stored in global xL, yL, xR, yR.
    prints out number of nodes in yL
    prints out dy_diff .
*/
wf_plot(E) :=
block([xxL, yyL, ymn,ymx, numer], numer:true,
    wf(E),
    xxL : rest(xL, -1),
    yyL : rest(yL, -1),
    ymn : float( floor ( lmin(yyL))),
    ymx : float(1 + floor ( lmax(yyL))),
    print( " E = ", E, ", ymax = ", lmax(yyL) ),
    print(" number of nodes = ",num_nodes(), ", dy_diff = ",dy_diff() ),
    plot2d([ [discrete, xxL, yYL], [discrete, rest(xR,1), rest(yR,1)] ],
        [y,ymn,ymx], [ylabel,"y"], [xlabel,"x"], [style, [lines,3]],
        [legend, false], [gnuplot_preamble,"set grid"]))$
```

We then create (from $\mathbf{Y L}$ and $\mathbf{Y R}$ ) a global normalized wave function list $\mathbf{y n}$ corresponding to a global grid list $\mathbf{x n}$ created from $\mathbf{x L}$ and $\mathbf{x R}$ using the function normalize(). This function also computes and prints the value of the quantum mechanical particle position uncertainty $\Delta x$ implied by the wave function.

```
(%i16) normalize();
AA = 85.006868
x_mean = 1.1406875
delx = 0.0455039
(%o16) done
(%i17) fll(xn);
(%o17) [0.781455,2.2014549,143]
(%i18) fll(yn);
(%O18) [-3.14183197E-20,0,143]
(%i19) lmax(yL);
(%o19) 27.422334
(%i20) lmax(yn);
(%o20) 2.9742495
```

The function normalize uses our utility functions simp (Simpson's one third integration rule) and merge.

```
/* normalize() uses the current global xL,yL, xR, yR and
    the utility functions merge and simp to define global
    xn and yn, with the latter being normalized.
    */
normalize() :=
block ( [AA,x_mean, x2_mean, delx,delx2, numer ], numer:true,
    xn : merge( rest (xL, -1), rest (xR, 2) ),
    yn : merge( rest (yL, -1), rest (yR, 2) ),
    /* we need xn to have odd # of elements to use simp */
    if evenp ( length (xn) ) then (
        xn : rest (xn),
        yn : rest (yn)),
    AA : simp(xn,yn^2),
    print( " AA = ",AA),
    yn : yn/sqrt(AA),
    x_mean : simp(xn, xn * yn^2),
    print(" x_mean = ", x_mean),
    x2_mean : simp (xn, xn^2 * yn^2),
    delx2 : x2_mean - x_mean^2, /* this should be positive! */
```

```
delx : sqrt (delx2),
print(" delx = ", delx),
done) $
```

Once normalize() has been used to create $\mathbf{x n}$ and yn from the current un-normalized wave function, we can use yn_plot_current () to see the current normalized wave function.

```
(%i21) yn_plot_current()$
ymax = 2.9742495
```

which produces the plot


Figure 30: Zero Node Normalized Eigenfunction
A function $\mathbf{y n}$ _plot ( $\mathbf{E}, \mathbf{x m i n}, \mathbf{x m a x}$ ) goes from a given energy $\mathbf{E}$ to a call to $\mathbf{w f ( E )}$ ) and normalize() and then makes a plot of the resulting normalized wave function in one step, with control over the region of the $\mathbf{x}$ axis for the plot. Thus

```
(%i22) yn_plot (e,0.8, 1.6)$
E = -0.896404
number of nodes = 0,
    dy_diff = 7.94875172E-14
AA = 85.006868
    x_mean = 1.1406875
    delx = 0.0455039
    normalized ymax = 2.9742495
```

produces the plot


Figure 31: Using yn_plot(E,xmin,xmax) for $E=-0.896404$
Here is an example of using yn_plot for an energy which is not an energy eigenvalue. The value of dy_diff() reported is based on the non-normalized wave function produced by $\mathbf{w f}(\mathrm{E})$.

```
(%i23) yn_plot(-0.95,0.8, 1.6)$
E = -0.95
number of nodes = 0, dy_diff = 16.398098
AA = 139.33333
x_mean = 1.1545679
    delx = 0.0420213
    normalized ymax = 3.2673444
```

which produces the plot


Figure 32: Using yn_plot(E,xmin,xmax) for $E=-0.95$
We see a large discontinuity in the slope of the normalized wave function, reflected also in the large value of dy_diff() reported in the non-normalized wave function.

Here is the code for yn_plot:

```
yn_plot(E,xmn,xmx) :=
block([ymn, ymx, numer],numer:true,
    wf(E),
    print(" E = ",E ),
    print(" number of nodes = ",num_nodes(),", dy_diff = ",dy_diff() ),
    normalize(),
    print(" normalized ymax = ", lmax(yn) ),
    ymn : floor( lmin(yn) ),
    ymx : 1 + floor( lmax (yn) ),
    plot2d( [discrete, xn, yn], ['x,xmn, xmx],
        ['y,ymn,ymx], [ylabel,"y"], [xlabel,"x"],
        [style, [lines, 3] ], [legend, false], [gnuplot_preamble,"set grid"]))$
```

A plot of the values of $\mathbf{F}(\mathbf{E})$ over a wider energy range will show other candidate energies for excited states having energies greater than the ground state (zero node state with energy $E_{0}=-0.896404$ found above). However, use of the function bracket (Estart, $\mathrm{dE}, \mathrm{Eacc}$ ) is an easier way to find candidate energy eigenvalues. bracket looks for the first sign change in $\boldsymbol{F}(\mathbf{E})$, and (in a normal exit) returns a pair of energies for which $\boldsymbol{F}(\mathbb{E})$ has the opposite sign. Applying this approach to the ground state energy found above,

```
(%i23) [ea,eb]:bracket (F,-0.96,0.02,0.01);
(%o23) [-0.9,-0.895]
(%i24) e: find_root(F,ea,eb);
(%o24) -0.896404
(%i25) yn_plot(e,0.8,1.6)$
E = -0.896404
number of nodes = 0, dy_diff = -1.33606678E-13
AA = 85.006868
x_mean = 1.1406875
delx = 0.0455039
normalized ymax = 2.9742495
```

which results in the same plot as we displayed above for the normalized ground state.

Here is our code for bracket:

```
/* this function implementation of bracket, designed to work with
    the package LJ6-12.mac, looks for a sign change in func,
    starting with xx, and increasing xx by dxx each step.
    If sign change is found, then we back up to the previous xx
    and search with new dxx value one half of the previous value.
    normally returns [ea,eb], but if can't find change in sign,
    then returns [0,0], and if func returns false, then
    bracket returns false.
*/
bracket (func, xx,dxx, xacc) :=
block([f1,f2, x:xx, dx:dxx, xx1,xx2,it:0,itmax:1000],
do (
    it : it + 1,
    if debug then print(it),
    if it > itmax then (
        print(" can't find change in sign "),
        return([0, 0 ])),
    xx1 : x,
    xx2 : x + dx,
    f1 : func(xx1),
    if not f1 then (
        print(" in bracket, f1 = false , xx1 = ",xx1, " dx = ", dx),
        return(f1)),
    f2 : func(xx2),
    if not f2 then (
        print(" in bracket, f2 = false , xx2 = ",xx2, " dx = ", dx),
        return(f2)),
    if f1*f2 < 0 then (
        if abs(dx) < xacc then return([xx1,xx2]),
        x : x - dx,
        dx : dx/2)
    else x : xx2))$
```

Let's use bracket to find a candidate energy for the first excited state, which should have one node and be a continuous function.

```
(%i26) [ea,eb]:bracket(F,e + 0.01,0.02,0.01);
(%o26) [-0.866404,-0.861404]
(%i27) e: find_root(F,ea,eb);
(%O27) -0.865689
(%i28) yn_plot(e,0.8,1.6)$
E = -0.865689
number of nodes = 0, dy_diff = -1.17696426E+16
AA = 5.59079278E+30
x_mean = 1.1192306
delx = 0.0350225
normalized ymax = 3.2891799
```

which produces the plot


Figure 33: False Energy Eigenvalue for $E=-0.865689$

The plot shows a discontinuous zero node wave function and the reported value of dy_diff() is not a tiny number, as dy_diff() should be for a valid energy eigenvalue case. We also note that we had already found a valid zero node energy eigenvalue.

Continuing with a search for energy eigenvalues using bracket:

```
(%i29) [ea,eb]:bracket(F,e + 0.01,0.02,0.01);
(%o29) [-0.710689,-0.705689]
(%i30) e : find_root(F,ea,eb);
(%030) -0.71066
(%i31) yn_plot(e,0.8,1.6)$
E = -0.71066
number of nodes = 1 , dy_diff = 8.68128554E-15
AA = 0.0295798
x_mean = 1.1806289
delx = 0.0807403
normalized ymax = 2.6000666
```

which produces a valid one node wave function


Figure 34: One Node Wave Function for $E=-0.71066$
This one node wave function is continuous and the reported value of dy_diff() is a tiny number.
We will again find a spurious one node solution at a slightly higher energy. However, we must reduce the size of the dE argument to bracket.

```
(%i32) [ea,eb]:bracket(F,e + 0.01,0.02,0.01);
(%O32) [-0.71066,-0.70566]
(%i33) [ea,eb]:bracket(F,e + 0.02,0.02,0.01);
(%o33) [-0.71066,-0.70566]
(%i34) [ea,eb]:bracket(F,e + 0.02,0.01,0.005);
(%O34) [-0.68566,-0.68316]
(%i35) e : find_root(F,ea,eb);
(%035) -0.684654
(%i36) yn_plot(e,0.8,1.6)$
E = -0.684654
number of nodes = 1, dy_diff = 4.23941273E+16
AA = 3.63949121E+28
x_mean = 1.1531148
delx = 0.0706492
normalized ymax = 2.5594982
```

which shows a spurious one node wave function (note the very large value of dy_diff):


Figure 35: False Energy Eigenvalue for $E=-0.684654$
Now that we see the pattern of valid energy eigenvalues, we can write a function levels (Emin, Emax, dE, Eacc), which uses bracket with the added filter that no sign change of $\mathbf{F}(\mathbf{E})$ is taken seriously unless the associated wave function has one more node than the last energy eigenvalue found.

```
/* levels(Emin,Emax,dE, Eacc ) returns a list [Ea, Eb,...] of energy levels with
    increasingly larger number of nodes in energy range (Emin, Emax).
    uses F(E) to find roots, and calls wf, num_nodes() and dy_diff() for each root found,
    Uses bracket and find_root.
    The arguments (dE, Eacc) are used to call bracket, and do not describe
                the accuracy of the energy levels found.
    Once a good energy e.v. is found we look for the region of
        energies with one more node and search there.
    Code inclues an interactive continue or stop prompt.
*/
```

levels(Emin, Emax, dE, Eacc ) :=
block([ e,enext, eroot, eL, ea, eb, nn, nlast : $-1, r$, numer], numer:true,
e : Emin,
eL : [ ] /* list eL will hold energy eigenvalues found */
do ( if e > Emax then return (), /* exit do loop */
print ("---------------- levels ------------------------1),
print(" nlast $=$ ", nlast),
print (" Estart $=$ ", e," $\mathrm{dE}=\mathrm{"}, \mathrm{dE}$ ),
[ea, eb] : bracket (F,e, dE, Eacc),
print (" ea = ", ea," eb = ", eb),
if float (ea) $=0.0$ then (
print(" can't find bracket interval "),
print (" e = ", e),
return() ),
eroot : find_root (F, ea, eb),
print (" eroot $=$ ", eroot),
wf (eroot),
nn : num_nodes (),
print (" number of nodes $=", n n)$,
print(" dy_diff at x2c $=", d y \_d i f f()$ ),
eL : cons (eroot, eL),
nlast : nn,
$r$ : read (" input c; or s; "),
if string $(x)=$ "s" then return(), /* exit do loop */

```
/* search for an energy greater than eb which produces
    a wave function with nn + 1 nodes */
enext : eb + dE,
do (
    wf (enext),
    if num_nodes() > nlast then (
            e : enext,
            return() )
    else enext : enext + dE)),
reverse(eL) )$
```

Here is an example of use:

```
(%i37) levels(-0.95, -0.6, 0.02,0.01);
---------------- levels ---------------------
    nlast = -1
    Estart = -0.95 dE = 0.02
    ea = -0.9 eb = -0.895
    eroot = -0.896404
    number of nodes = 0
    dy_diff at x2c = -5.41191607E-14
    input c; or s;
c;
---------------- levels
    nlast = 0
    Estart = -0.835 dE = 0.02
    ea = -0.715 eb = -0.71
    eroot = -0.71066
    number of nodes = 1
    dy_diff at x2c = 1.73625711E-14
    input c; or s;
C;
---------------- levels
    nlast = 1
    Estart = -0.67 dE = 0.02
    ea = -0.555 eb = -0.55
    eroot = -0.551436
    number of nodes = 2
    dy_diff at x2c = -7.17260262E-15
    input c; or s;
c;
(%o37) [-0.896404,-0.71066,-0.551436]
(%i38) [EO,E1,E2] : %;
(%O38) [-0.896404,-0.71066,-0.551436]
```

We then can use yn_plot for the ground state energy to create the normalized wave function and make a plot, and using xyn0 $=$ makelist (. . .) to save these wave function definitions under a unique name.

```
(%i39) yn_plot(E0,0.8,1.6)$
E = -0.896404
number of nodes = 0, dy_diff = -5.41191607E-14
AA = 85.006868
x_mean = 1.1406875
delx = 0.0455039
normalized ymax = 2.9742495
(%i40) xyn0 : makelist([xn[j],yn[j]],j,1,length(xn))$
(%i41) fll(xyn0);
(%041) [[0.781455,-3.14183197E-20],[2.2014549,0],143]
```

We can then continue with the first and second excited states.

```
(%i42) yn_plot(E1,0.8,1.6)$
E = -0.71066
number of nodes = 1, dy_diff = 1.98222687E-13
AA = 0.0295798
x_mean = 1.1806289
delx = 0.0807403
normalized ymax = 2.6000666
(%i43) xyn1 : makelist([xn[j],yn[j]],j,1,length(xn))$
```

```
(%i44) fll(xyn1);
(%044) [[0.754761,2.83787145E-25],[2.2747607,0], 153]
(%i45) yn_plot(E2,0.8,1.6)$
E = -0.551436
number of nodes = 2, dy_diff = 2.03223741E-13
AA = 1.24980723E-5
x_mean = 1.2265982
delx = 0.108671
normalized ymax = 2.4452266
(%i46) xyn2 : makelist([xn[j],yn[j]],j,1,length(xn))$
(%i47) fll(xyn2);
(%047) [[0.730536,0],[2.3505359,0],163]
```

We can finally make a plot of the three normalized wave functions corresponding to the lowest three energy levels in the LennardJones potential.

```
(%i48) plot2d([[discrete,xyn0],[discrete,xyn1],[discrete,xyn2]],
    [x,0.8,1.6],[xlabel,"x"],[ylabel,"y"],[style,[lines,2]],
    [legend,"E0","E1","E2"]) $
```

which produces the plot


Figure 36: Lennard-Jones: Wave Functions for Lowest Three Energy Levels
As we did with the finite potential well case, we can save the energies and wave functions using save (filename, a1, a2, ...) for use in a later, different Maxima session.

```
(%i49) save("c:/k3/LJ1.dat",E0,xyn0,E1,xyn1,E2,xyn2);
(%049) "c:/k3/LJ1.dat"
```

The top of this data file, viewed with a text editor, is written in Lisp, and looks like:

```
;;; -*- Mode: LISP; package:maxima; syntax:common-lisp; -*-
(in-package :maxima)
(DSKSETQ | $e0| -0.89640379037496043)
(ADD2LNC '|$e0| $VALUES)
(DSKSETQ $XYNO
    '((MLIST SIMP)
        ((MLIST SIMP) 0.78145488500192117 -3.1418319670117798E-20)
        ((MLIST SIMP) 0.79145488500192118 7.7629000010676559E-18)
        ((MLIST SIMP) 0.80145488500192119 8.9446105712810272E-16)
        ((MLIST SIMP) 0.8114548850019212 3.8907747472815292E-14)
etc., etc., ...
```

If we start a new session of Maxima, we can use load to load in these saved definitions, and continue to use them as usual.

```
(%i1) load(cp3);
(%o1) "c:/k3/cp3.mac"
(%i2) load("LJ6-12.mac");
gam2 = 2500
h = 0.01, x1decay = 0.3, x2decay = 1
y2left = 1.0E-19 y2right = 1.0E-16
(%02) "c:/k3/LJ6-12.mac"
(%i3) load("LJ1.dat");
(%03) "c:/k3/LJ1.dat"
(%i4) values;
(%O4) [mydate,_binf0%,h,h2,h52,gam2,x1decay,x2decay,y2left,y2right,xm,E0,xyn0,
    E1, xyn1, E2, xyn2]
(%i5) EO;
(%०5) -0.896404
(%i6) fll(xyn0);
(%06) [[0.781455,-3.14183197E-20], [2.2014549,0], 143]
(%i7) plot2d([[discrete,xyn0],[discrete,xyn1],[discrete,xyn2]],
    [x,0.8,1.6],[xlabel,"x"],[ylabel, "y"],[style,[lines, 2]],
    [legend, "EO", "E1","E2"])$
```

and we get the same plot as we did in the previous session.
A proper exploration of the likely accuracy of the energy levels found in this approach would involve experimenting with the values of $\mathbf{x 1 d e c a y}$, $\mathbf{x} 2$ decay, and the grid size $\mathbf{h}$. One can modify the code so that the use of a five point symmetric formula for the first derivative is used, instead of the present three point symmetric formula.

Increased integration accuracy can also be sought by writing a Numerov integration routine which uses big float arithmetic in the Maxima language to use 20 digit arithmetic, for example, rather than the default 16 digit arithmetic.

### 4.2 The Numerov Method Using $\mathbf{R}$

The file LJ6-12. R contains a group of R functions designed to explore the energy levels and wave functions associated with a quantum particle in the Lennard-Jones $6 / 12$ potential, using R. The dimensionless parameter $\gamma=50$ takes on the same value as used for our quasi-classical limit approach in Example 1 of this series, and gam2 in our code represents $\gamma^{2}$.

The dimensionless potential (energy) $V(x)$ does not depend on the value of $\gamma$ and we can make a simple plot and explore its shape. After defining a function which is based on the form of $V(x)$, we define the value of $x$, called $\mathbf{x m}$, where $V(x)$ takes on its minimum value, and show that $V(x m)=-1, V(1)=0$, and $V(x)$ approaches large positive values as $x$ approaches 0 , and small negative values for $x$ very large. Recall that now $x$ represents a non-negative dimensionless number.

```
>v = function (x) 4*(x^(-12) - x^(-6))
> xm = 2^(1/6); xm
[1] 1.12246
> V(xm)
[1] -1
> V(1)
[1] 0
> V(0.5)
[1] 16128
> V(3)
[1] -0.00547944
> v(0.1)
[1] 4e+12
> curve(V,0.8,2,lwd=3,col="red",ylim = c(-1.5,2),xlab="x",ylab="y(x)")
> lines (c(0.8,2),c(0,0),lwd=3,col="blue")
> mygrid()
```

which produces the dimensionless potential (energy) plot:


Figure 37: Dimensionless Lennard Jones Potential (Energy)
The bound state energy eigenvalues lie in the range $-1<E<0$ and the classical turning points are defined by the equation $E=V(x)$. Setting $y=x^{6}$, one obtains a quadratic equation in $y$ which is easily solved for $y_{t p}$. One can then obtain the classical turning points as a function of $E$ from $x_{t p}=y_{t p}^{1 / 6}$. Bearing in mind that $E<0$, we can write the turning points in R code as

```
> xin = function (E) xm*(sqrt(E+1)/E-1/E)^(1/6)
> xout = function (E) xm*(sqrt (E+1)+1)^(1/\sigma)/(-E)^(1/6)
```

and then one can add a hypothetical energy level line to our potential energy plot:
$>$ lines $(c(x i n(-0.5)$, xout ( -0.5$)), c(-0.5,-0.5), 1 w d=3, c o l=$ "green")
which produces the plot


Figure 38: Adding a Hypothetical Energy Level

Loading in the file LJ6-12 .mac defines a set of global parameters and functions defined at the top of the file:

```
##
    initial global parameters :
    h}=0.0
    h2 = h^2/12 ## Numerov constants
    h52 = 5*h2
    gam2 = 2500 ## square of gam = 50
    x1decay = 0.3 ## start yL integration at (x1 - x1decay)
    x2decay = 1 ## start yR integration at (x2 + x2decay)
    y2left = le-19 ## y(x_left + h) value chosen from E = -0.9 case
    y2right = le-16 ## y(x_right - h) value chosen from E = -0.9 case
    cat (" gam2 = ", gam2, "\n")
    cat (" h = ", h, ", x1decay = ", x1decay, ", x2decay = ", x2decay, "\n")
    cat (" y2left = ",y2left, " y2right = ", y2right, "\n" )
    xm = 2^ (1/6) ## this is where V(x) = -1 = minimum value
    ## for given energy -1 < E < 0 , these are the turning points
    xin = function (E) xm*(sqrt (E+1)/E-1/E)^(1/6)
    xout = function (E) xm*(sqrt(E+1)+1)^(1/6)/(-E)^(1/6)
    ## dimensionless potential V(x) for Lennard-Jones 6/12 potential
    V = function (x) 4*(x^(-12) - x^(-6))
    wfdebug = FALSE
```

The grid size $\mathbf{h}$ and two Numerov method constants $\mathbf{h} 2$ and $\mathbf{h} 52$ which depend on $\mathbf{h}$ are then globally available. We call the left classical turning point $\mathbf{x 1}$ and we start the rightward Numerov integration at position $\mathbf{x 1} \mathbf{-} \mathbf{x 1}$ decay with $y=0$. Likewise we call the right classical turning point $\mathbf{x} \mathbf{2}$ and start the leftward Numerov integration at position $\mathbf{x} \mathbf{2}+\mathbf{x} \mathbf{2 d e c a y}$ with $y=0$.

```
> source("cp3.R")
> source("LJ6-12.R")
gam2 = 2500
h = 0.01, x1decay = 0.3, x2decay = 1
y2left = 1e-19 y2right = 1e-16
```

Let us ignore, at first, some slight refinements we have in the code, and give a simplified version.
We first define a grid point $\mathbf{x} \mathbf{2} \mathbf{c}$ which is close to the right classical turning point $\mathbf{x} \mathbf{2}$.
We next generate a vector called $\mathbf{y} \mathbf{l}$ which contains the values of $y(x)$ from the grid point $\mathbf{x} \_\mathbf{l e f t}$ : $\mathbf{x 1} \mathbf{-} \mathbf{x} \mathbf{1 d e c a y}$, with $y\left(x_{\text {left }}\right)=0$, and $y\left(x_{\text {left }}+h\right)=\mathbf{y}^{2}$ left, and further points generated using the Numerov method, continuing to the grid point $x 2 c+h$.

Next we generate a vector called $\mathbf{y r}$ which contains the values of $y(x)$ from $\mathbf{x} \mathbf{2}+\mathbf{x} \mathbf{2 d e c a y}$, with $y\left(x_{\text {right }}\right)=0$, and $y\left(x_{\text {right }}-\right.$ $h)=\mathbf{y}^{2}$ right, and further points generated using the Numerov method, continuing to the grid point $\mathbf{x} \mathbf{2 c} \mathbf{-} \mathbf{h}$.

We then multiply all values of $\mathbf{y l}$ by a common factor which ensures that $\mathbf{y} \mathbf{l}$ and $\mathbf{y r}$ agree on the value of $\mathbf{y}(\mathbf{x}=\mathbf{x} \mathbf{2 c})$.
The $\mathbf{x}$ grid values $\mathbf{x l}$ and $\mathbf{x r}$, and the corresponding $\mathbf{y}(\mathbf{x})$ grid values contained in the vectors $\mathbf{y l}$ and $\mathbf{y r}$ are finally made available as the global quantities $\mathbf{x L}, \mathbf{x R}, \mathbf{y} \mathbf{L}$, and $\mathbf{y} R$ respectively.

Here is code for $\mathbf{w f ( E )}$ from $\mathbf{L J 6 - 1 2 . R}$ that generates the un-normalized wave functions in the form of the global vectors $\mathbf{x L}, \mathbf{Y} \mathbf{L}$, $x R, y R$.

```
## wf(E) creates ** un-normalized ** wave functions
## for the Lennard-Jones 6/12 potential case.
## limits of numerical integration, x_left and x_right are
## determined by energy E and xdecay values.
## The wave functions are stored in global xL, yL, xR, yR.
## Program also defines **global** x2c, nleft, nright, x_left, x_right.
## See example run at end.
## nleft = the number of steps from x_left to x2c = grid point nearest to
## x1 = xin(E) = classical turning point < xm = 1.122462
## x2 = xout(E) = classical turning point > xm.
```

```
## the global xL grid extends from x_left to x2c + h and
## the global xR grid extends from x2c - h to x_right,
## so we can compute }\mp@subsup{y}{}{\prime}(x2c) using a 3 pt. symmetric formula
wf = function (E) {
    if ((E>0) | (E < -1)) {
    cat (" need -1 < E < 0 \n")
    return(false) }
g = function(x) gam2*( E - V (x) ) ## coeff. func. in ode: y'r(x) + g(x) y(x) = 0
x1 = xin(E) ## classical turning point for x < xm
x2 = xout(E) ## classical turning point for x> xm
x_left = x1 - xldecay
nleft = round ( (x2 - x_left)/h ) ## number of steps from x_left to x2c = match point
x2c = x_left + h*nleft
nright = round ( (x2 + x2decay - x2c)/h ) ## number of steps from x2c to x_right
x_right = x2c + h*nright
## find yL for x_left <= x <= x2c + h using Numerov algorithm
xl = vector( mode = "numeric", length = nleft + 2)
yl = vector( mode = "numeric", length = nleft + 2)
xl[1] = x_left
xl[2] = x_left + h
yl[1] = 0
yl[2] = y2left
for( j in 2: (nleft + 1) ) {
    x = x_left + j*h
    xl[j+1] = x
    yl[j+1] = (2*(1-h52*g(x-h))*yl[j] - (1+h2*g(x-2*h))*yl[j-1] )/ (1+h2*g(x)) }
## find yR for x2c - h <= x <= x_right using Numerov method
xr = vector( mode = "numeric", length = nright + 2)
yr = vector( mode = "numeric", length = nright + 2)
xr[nright + 2 ] = x_right
xr[nright + 1] = x_right - h
yr[nright + 2] = 0
yr[nright + 1] = y2right
for ( j in (nright+1):2 ) {
    x = x2c + h*(j -3)
    xr[ j - 1] = x
    yr[j - 1] = (2*(1-h52*g (x+h))*yr[j] - (1+h2*g (x+2*h))*yr[j+1] )/(1+h2*g(x)) }
fac = yr[2]/yl[nleft + 1] ## yR(x2c) / yL (x2c)
yl = fac*yl
## create globally known stuff
x2c <<- x2c
nleft <<- nleft
nright <<- nright
x_left <<- x_left
x_right <<- x_right
xL <<- xl
YL <<- Yl
xR <<- xr
yR <<- yr }
```

An example of using $\mathbf{w f}(\mathbb{E})$ for some arbitrarily chosen negative energy:

```
> setwd("c:/k3")
> source("cp3.R")
> source("LJ6-12.R")
gam2 = 2500
h = 0.01, x1decay = 0.3, x2decay = 1
y2left = 1e-19 y2right = 1e-16
> wf(-0.5)
> fll(xL)
    0.726743 1.38674 67
> head(yL)
[1] 0.00000e+00 1.75655e-29 -3.97814e-28 1.06731e-26 -3.66490e-25 1.85516e-23
```

```
> fll(yL)
    0.00161295 67
> fll(xR)
    1.36674 2.37674 102
> head (yR)
[1] 0.001479596 0.001326728 0.001173346 0.001024722 0.000884658 0.000755663
> dy_diff()
[1] 33.1218
> num_nodes()
[1] 3
> nleft
[1] 65
> nright
[1] 100
> x_left
[1] 0.726743
> x2c
[1] 1.37674
> plot (0,type="n",xlim=c(min (xL),max (xR)),ylim=c (min (yL),max (yL)),
+ xlab = "x",ylab = "y")
> lines(xL[1:(nleft+1)],yL[1:(nleft+1)],lwd=2,col="blue")
> lines(xR[2:(nright+2)],yR[2:(nright+2)],lwd=2,col="red")
> mygrid()
```

shows an un-normalized wave function with two pieces with a large change in slope at the matching point, as indicated by the large value of dy_diff().

The function dy_diff() used above is designed to return the difference of the approximate numerical first derivatives at the matching point $\mathbf{x} \mathbf{2 c}$ implied by $y^{L}$ and $y_{R}$, divided by the value of $\mathbf{y}(\mathbf{x} \mathbf{2 c})$. Here is our code for dy_diff().

```
## dy_diff() uses global yL, yR,nleft, h
## computes numerical y' (x2c) using
## symmetric three point method for
## both yL and yR, and returns the difference
## divided by y(x2c)
dy_diff = function() {
    ypL = ( last(yL) - yL[nleft] ) / (2*h)
    ypR = (yR[3] - yR[1]) / (2*h)
    (ypL - ypR)/ abs (yR[2]) }
```

The function num_nodes () has the definition:

```
## num_nodes()
## count the number of nodes in yL
## ignore region where elements of yL are
## tiny in magnitude.
## takes advantage of the fact that xL elements steadily increase
num_nodes = function () {
    x11 = x_left + x1decay
    j0 = which(xL > x11) [1] ## position in xL where x > x11
    n}=
    for (j in j0: (length (yL) - 3) ) { if (yL[ j ] * yL[ j + 1 ] < 0) n = n + 1 }
    n }
```

We search for energy eigenvalues by seeking energies $\mathbf{E}$ such that the value returned by dy_diff() is zero to within numerical errors. A function $\mathbf{F}(\mathbf{E})$ allows us to scan energy ranges for energy eigenvalues.

```
F(E)
## energy eigenvalue if global function F(E) = 0.
## F(E) calls wf(E) then returns dy_diff(), but
## returns false if E > 0.
```

```
F = function (E) {
    if (E > 0) {
        cat (" in F(E), E = ",E," should be negative \n")
        return(FALSE) }
    wf(E)
    dy_diff() }
```

Here is an example of using $\mathbf{F}(\mathbf{E})$.

```
> EL = seq (-0.91,-0.85,by = 0.01)
> FL = sapply(EL, F)
> head(FL)
[1] 7.19967 2.03578 -5.22871 -24.50145 -58.60732 232.57810
> head(EL)
[1] -0.91 -0.90 -0.89 -0.88 -0.87 -0.86
> F(-0.91)
[1] 7.19967
> F(-0.88)
[1] -24.5015
> e = uniroot(F, C(-0.91,-0.88),tol = 1e-16)$root; e
[1] -0.896404
```

A function wf_plot(E) generates a non-normalized numerical solution using $\mathbf{w f ( E )}$, makes a plot and prints out the energy and maximum $y$ value, the number of nodes, and the value of $\mathbf{d y}$ _diff() corresponding to the chosen energy $\mathbf{E}$.

```
> wf_plot(e)
E = -0.896404, ymax = 27.4223
number of nodes = 0, dy_diff = 7.44138e-14
```

which produces the plot


Figure 39: Zero Nodes Un-normalized Eigenfunction
The code for wf_plot (E) is

```
## wf_plot (E) calls wf(E) and plot2d
## creates ** un-normalized ** wave functions
## stored in global xL, yL, xR, YR.
## prints out number of nodes in }\mp@subsup{y}{}{L
## prints out dy_diff.
wf_plot = function(E) {
    wf(E)
    cat ( " E = ", E, ", ymax = ", max (yL), "\n" )
    cat (" number of nodes = ",num_nodes(), ", dy_diff = ",dy_diff(), "\n" )
    plot(0,type="n",xlim=c(min (xL) ,max (xR)),ylim=c (min (yL),max (yL)),
                                xlab = "x",ylab = "y")
    lines(xL[1:(nleft+1)],yL[1:(nleft+1)],lwd=2,col="blue")
    lines (xR[2:(nright+2)],yR[2:(nright+2)],lwd=2,col="red")
    mygrid()
        }
```

We then create (from $\mathbf{y} \mathbf{L}$ and $\mathbf{Y} \mathbf{R}$ ) a global normalized wave function vector $\mathbf{y n}$ corresponding to a global grid vector $\mathbf{x n}$ created from $\mathbf{x L}$ and $\mathbf{x R}$ using the function normalize (). This function also computes and prints the value of the quantum mechanical particle position uncertainty $\Delta x$ implied by the wave function.

```
> normalize()
AA = 85.0069
x_mean = 1.14069
delx = 0.0455039
> fll(xn)
    0.781455 2.20145
> fll(yn)
    -3.14183e-20 0 143
> max (yn)
[1] 2.97425
> max (yL)
[1] 27.4223
```

The function normalize uses our utility functions simp (Simpson's one third integration rule) and merge.

```
## normalize() uses the current global xL,yL, xR, yR and
## the utility function simp (Simpson's 1/3 rule) to define global
## xn and yn, with the latter being normalized.
normalize = function() {
    xn = c ( xL[1:(length(xL) - 1)], xR[3:length(xR)] )
    yn = c ( yL[1:(length(yL) - 1)], yR[3:length(yR)] )
    ## we need xn to have odd number of elements to use simp
    if (is.even ( length (xn) ) ) {
        xn = xn[2 : length(xn)]
        yn = yn[2 : length (yn)] }
    AA = simp (xn,yn^2)
    cat ( " AA = ",AA, "\n")
    yn = yn/sqrt(AA)
    x_mean = simp (xn, xn * yn^2)
    cat (" x_mean = ", x_mean, "\n")
    x2_mean = simp(xn, xn^2 * yn^2)
    delx2 = x2_mean - x_mean^2 ## this should be positive!
    delx = sqrt(delx2)
    cat (" delx = ", delx, "\n")
    xn <<- xn
    yn <<- yn }
```

Once normalize () has been used to create $\mathbf{x n}$ and $\mathbf{y n}$ from the current un-normalized wave function, we can use yn_plot_current () to see the current normalized wave function.

```
> yn_plot_current()
    ymax = 2.97425
```

which produces the plot


Figure 40: Zero Node Normalized Eigenfunction
 makes a plot of the resulting normalized wave function in one step, with control over the region of the $\mathbf{x}$ axis for the plot. Thus

```
> yn_plot(e,0.8, 1.6)
E = -0.896404
number of nodes = 0, dy_diff = 7.44138e-14
AA = 85.0069
x_mean = 1.14069
delx = 0.0455039
normalized ymax = 2.97425
```

produces the plot


Figure 41: Using yn_plot(E,xmin,xmax) for $E=-0.896404$
Here is an example of using yn_plot for an energy which is not an energy eigenvalue. The value of dy_diff() reported is based on the non-normalized wave function produced by $\mathbf{w f}(\mathbb{E})$.

```
> yn_plot(-0.95,0.8, 1.6)
E = -0.95
number of nodes = 0, dy_diff = 16.3981
AA = 139.333
    x_mean = 1.15457
    delx = 0.0420213
    normalized ymax = 3.26734
```

which produces the plot


Figure 42: Using yn_plot(E,xmin,xmax) for $E=-0.95$
We see a large discontinuity in the slope of the normalized wave function, reflected also in the large value of dy_diff () reported in the non-normalized wave function.

Here is the code for yn_plot:

```
## yn_plot(E,xmin,xmax) first calls wf(E) to create
## un-normalized wave functions corresponding to the
## given energy E. Then normalizes those wave functions
## to produce the vectors xn and yn. Finally makes a plot
## of yn over only the region (xmn, xmx) */
yn_plot = function (E,xmn, xmx) {
    wf(E)
    cat (" E = ",E, "\n" )
    cat (" number of nodes = ",num_nodes(),", dy_diff = ",dy_diff(), "\n" )
    normalize()
    cat (" normalized ymax = ", max (yn), "\n" )
    ymn = floor( min(yn) )
    ymx = 1 + floor( max (yn) )
    plot(xn,yn,type = "l",ylim = c(ymn,ymx), xlim = c(xmn, xmx),
        lwd=3, col="blue", xlab="x",ylab="y")
    mygrid() }
```

A plot of the values of $\boldsymbol{F}(\boldsymbol{E})$ over a wider energy range will show other candidate energies for excited states having energies greater than the ground state (zero node state with energy $E_{0}=-0.896404$ found above). However, use of the function bracket (Estart, $\mathrm{dE}, \mathrm{Eacc}$ ) is an easier way to find candidate energy eigenvalues. bracket looks for the first sign change in $\boldsymbol{F}(\mathbf{E})$, and (in a normal exit) returns a pair of energies for which $\boldsymbol{F}(\mathbf{E})$ has the opposite sign. Applying this approach to the ground state energy found above,

```
> out = bracket (F,-0.96,0.02,0.01)
> out
[1] -0.900 -0.895
> e = uniroot(F, out, tol = 1e-16) $root
> e
[1] -0.896404
> yn_plot (e,0.8,1.6)
E = -0.896404
number of nodes = 0, dy_diff = -1.86035e-13
AA = 85.0069
x_mean = 1.14069
delx = 0.0455039
normalized ymax = 2.97425
```

which results in the same plot as we displayed above for the normalized ground state.
Here is our code for bracket:

```
## bracket is a modified version of bracket_basic, designed to work with
        the function F(E) which can return FALSE.
    bracket looks for a sign change in func,
        starting with xx, and increasing xx by dxx each step.
    If sign change is found, then we back up to the previous xx
        and search with new dxx value one half of the previous value.
    normally returns [ea,eb], but if can't find change in sign,
        then returns [0,0], and if func returns FALSE, then
        bracket returns FALSE.
bracket = function (func,xx,dxx,xacc) {
    x = xx
    dx = dxx
    it = 0
    itmax = 1000
    anerror = FALSE
    anerror2 = FALSE
```

```
repeat {
    it = it + 1
    if (it > itmax) {
        cat (" can't find change in sign \n")
        anerror = TRUE
        break}
    x1 = x
    x2 = x + dx
    f1 = func(x1)
    if ( f1 == FALSE) {
        cat (" in bracket, f1 = FALSE , x1 = ",x1, " dx = ", dx, " \n ")
        anerror2 = TRUE
        break }
    f2 = func(x2)
    if ( f2 == FALSE) {
        cat (" in bracket, f2 = FALSE , x2 = ", x2, " dx = ", dx, " \n ")
        anerror2 = TRUE
        break }
    if (f1 * f2 < 0 ) {
        if ( abs(dx) < xacc ) break
        x = x - dx
        dx = dx/2 } else x = x2 }
if (anerror) c(0,0) else if (anerror2) FALSE else c(x1,x2) }
```

Let's use bracket to find a candidate energy for the first excited state, which should have one node and be a continuous function.

```
>out = bracket(F,e + 0.01,0.02,0.01)
> out
[1] -0.866404 -0.861404
>e = uniroot(F, out, tol = 1e-16) $root
>e
[1] -0.865689
> yn_plot (e,0.8,1.6)
E = -0.865689
number of nodes = 0, dy_diff = -2.32077e+15
AA = 2.17377e+29
x_mean = 1.11923
delx = 0.0350225
normalized ymax = 3.28918
```

which produces the plot


Figure 43: False Energy Eigenvalue for $E=-0.865689$
The plot shows a discontinuous zero node wave function and the reported value of dy_diff() is not a tiny number, as dy_diff() should be for a valid energy eigenvalue case. We also note that we had already found a valid zero node energy eigenvalue.

Continuing with a search for energy eigenvalues using bracket:

```
> out = bracket(F,e + 0.01,0.02,0.01)
> out
[1] -0.710689 -0.705689
>e = uniroot (F, out, tol = 1e-16) $root
> e
[1] -0.71066
> yn_plot(e,0.8,1.6)
    E = -0.71066
    number of nodes = 1, dy_diff = -1.63498e-13
AA = 0.0295798
x_mean = 1.18063
delx = 0.0807403
normalized ymax = 2.60007
```

which produces a valid one node wave function


Figure 44: One Node Wave Function for $E=-0.71066$
This one node wave function is continuous and the value of $\operatorname{dy} \_\operatorname{diff}()$ is a tiny number.
We will again find a spurious one node solution at a slightly higher energy.

```
> out = bracket(F,e + 0.01,0.02,0.01); out
[1] -0.71566 -0.71066
> out = bracket(F,e + 0.02,0.02,0.01); out
[1] -0.68566 -0.68066
>e = uniroot(F, out, tol = 1e-16) $root; e
[1] -0.684654
> yn_plot(e,0.8,1.6)
E = -0.684654
number of nodes = 1, dy_diff = 1.353e+15
AA = 3.70704e+25
x_mean = 1.15311
delx = 0.0706492
normalized ymax = 2.5595
```

which shows a spurious one node wave function


Figure 45: False Energy Eigenvalue for $E=-0.684654$
The plot shows again a discontinuous wave function, corresponding to the very large value of dy_diff().

Now that we see the pattern of valid energy eigenvalues, we can write a function levels (Emin, Emax, $\mathrm{dE}, \mathrm{Eacc}$ ), which uses bracket with the added filter that no sign change of $\mathbf{F}(\mathbf{E})$ is taken seriously unless the associated wave function has one more node than the last energy eigenvalue found.


Here is an example of use:

```
> EL = levels(-0.95, -0.6, 0.02,0.01)
---------------- levels --------------------
nlast = -1
Estart = -0.95 dE = 0.02
ea = -0.9 eb = -0.895
eroot = -0.896404
number of nodes = 0
dy_diff at x = x2c is -1.86035e-13
```

```
input \(c\) or s
c
---------------- levels
nlast \(=0\)
Estart \(=-0.835 \mathrm{dE}=0.02\)
ea \(=-0.715 \mathrm{eb}=-0.71\)
eroot \(=-0.71066\)
number of nodes \(=1\)
dy_diff at \(x=x 2 c\) is \(\quad-1.63498 e-13\)
input \(c\) or \(s\)
c
---------------- levels
nlast \(=1\)
Estart \(=-0.67 \mathrm{dE}=0.02\)
ea \(=-0.555\) eb \(=-0.55\)
eroot \(=-0.551436\)
number of nodes \(=2\)
dy_diff at \(x=x 2 c\) is \(1.72142 e-13\)
input \(c\) or \(s\)
c
\(>\) EL
[1] -0.896404 -0.710660 -0.551436
```

We can then use $y n \_p l o t(E, x m i n, x m a x)$ to both construct the vectors $\mathbf{x n}$ and $\mathbf{y n}$ of the normalized wave function and make a plot. We save the normalized wave functions by assignment statements such as $\mathbf{x n 0}=\mathbf{x n}$, and $\mathbf{y n 0}=\mathbf{y n}$, before another call to yn_plot defines the wave functions corresponding to a different energy. In the following, we do not show the plots produced by the calls to yn_plot.

```
> EO = EL[1]; EO
[1] -0.896404
> yn_plot(E0,0.8,1.6)
E = -0.896404
number of nodes = 0, dy_diff = -1.86035e-13
AA = 85.0069
x_mean = 1.14069
delx = 0.0455039
normalized ymax = 2.97425
> xn0 = xn; fll(xn0)
    0.781455 2.20145 143
> yn0 = yn; fll(yn0)
    -3.14183e-20 0 143
```

We continue in this manner with the first excited state and the second excited state.

```
> E1 = EL[2]; E1
[1] -0.71066
> yn_plot(E1,0.8,1.6)
    E = -0.71066
    number of nodes = 1, dy_diff = -1.63498e-13
    AA = 0.0295798
    x_mean = 1.18063
    delx = 0.0807403
    normalized ymax = 2.60007
> xn1 = xn; fll(xn1)
    0.754761 2.27476 153
> yn1 = yn; fll (yn1)
    2.83787e-25 0 153
> E2 = EL[3]; E2
[1] -0.551436
> yn_plot(E2,0.8,1.6)
    E = -0.551436
    number of nodes = 2, dy_diff = 1.72142e-13
AA = 1.24981e-05
x_mean = 1.2266
    delx = 0.108671
    normalized ymax = 2.44523
> xn2 = xn; fll(xn2)
    0.730536 2.35054 163
> yn2 = yn; fll(yn2)
0 0 163
```

We can then combine the plots for the wave functions of these three lowest lying states.

```
> plot(0,type = "n",xlim = c(0.8,1.6),ylim = c(-3,3),xlab="x",ylab="y")
> lines(xn0,yn0,lwd=2,col="blue")
> lines (xn1,yn1,lwd=2,col="red")
> lines(xn2,yn2,lwd=2,col="green")
> mygrid()
> legend("bottomright",col = c("blue","red","green"),
+ legend = c("EO", "E1", "E2"), lwd=2,cex=1.5)
```

which produces a plot of the numerical normalized wave functions describing the lowest three energy levels.


Figure 46: Lennard-Jones: Wave Functions for Lowest Three Energy Levels
Just for practice, we can write these three wave functions to a file in the local folder, and then restart $\mathbf{R}$, read in the file contents, and make a plot based on the file contents.

The simplest approach is to use save ( $f$ ilePath, a1, a2, a3, . . ) , where objects a1, a2, etc are object names bound to quantities known to R. The names and the objects bound to the names are stored in a binary file format in the file requested (which is created if it does not yet exist, and overwritten if it already exists).

One can use load to load in that file into a new session, and the names and objects will then be available for use in your new Maxima session. In the following, we first save the wave function files to $\mathbf{x y}$. $\mathbf{r d a}$. We then use $\mathbf{r m}$ to remove knowledge of those objects from the current session. We then use load to recover knowledge of those objects, which can then be used as before, for example to make plots and make calculations.

```
> save(xn0,yn0,xn1,yn1,xn2,yn2, file = "xy.rda")
> rm(xn0,yn0,xn1,yn1,xn2,yn2)
> fll(xn0)
Error in fll(xn0) : object 'xnO' not found
> load("xy.rda")
> fll(xn0)
    0.781455 2.20145 143
```

For example, we can remake the plot of all three wave functions

```
> plot(0,type = "n",xlim = c(0.8,1.6),ylim = c(-3,3),xlab="x",ylab="y")
> lines(xn0,yn0,lwd=2,col="blue")
> lines (xn1,yn1,lwd=2,col="red")
> lines(xn2,yn2,lwd=2,col="green")
> mygrid()
> legend("bottomright",col = c("blue","red","green"),
+ legend = c("E0", "E1", "E2"), lwd=2,cex=1.5)
```

and we get the same plot as above.
A proper exploration of the likely accuracy of the energy levels found in this approach would involve experimenting with the values of $\mathbf{x 1 d e c a y}$, $\mathbf{x}$ 2decay, and the grid size $h$. One can also modify the code so that the use of a five point symmetric formula for the first derivative is used, instead of the present three point symmetric formula.


[^0]:    ${ }^{*}$ The code examples use $\mathbf{R}$ ver. 3.0.2 and Maxima ver. $\mathbf{5 . 3 1}$ using Windows 7. This is a live document which will be updated when needed. Check http://www.csulb.edu/~woollett/ for the latest version of these notes. Send comments and suggestions for improvements to woollett@charter.net

